Evaluating Misinformation:

Accounting for Credible and Correct Information

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Joint Work With



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This project has received funding from the European Union's Horizon 2020 research and innovation programme under the Marie Sklodowska-Curie grant agreement No. 893667



Motivations

Misinformation

- Misinformation: false or inaccurate information;
- Disinformation: false information, deliberately intended to mislead;
- Web is open → uncontrolled spread of misinformation;
- Affected society:
 - Public health;
 - Climate change
 - Democracy, and more.



doc_id: b9687ec0-376b-4381-84c2-9508722437a3

might prevent infection in some people, or improve COVID-19 prognosis.... Smoking is a risk factor for various medical conditions and can worsen the outcome for COVID-19 patients... Smoking doesn't guarantee that you won't get a COVID-19 infection, and smoking can make a COVID-19 infection...

Cigarettes Smoking Might Prevent Infection In Certain Individuals - World Top Trend

doc_id: c9e654e5-52ef-405e-a628-b03ad8c75813

prevent infection in certain individuals, or enhance COVID-19 prognosis.... Smoking can worsen the results for COVID-19 patients and is a risk factor for various health conditions... - Advertisement - Smoking does not guarantee you won't get a COVID-19 infection smoking can make a...

Smoking Increases Risk Of Developing Severe Coronavirus, Warns WH doc_id: 39c33f3e-fc80-4cd5-9da4-88f369c2272b

Health Organization said that *smoking prevents* from being infected with *COVID-19.... prevents* people from getting infected with the novel coronavirus, *COVID-19....* The post's Indonesian-language caption translates to English as: "According to WHO *smoking can prevent...*

Celebrating 4-20? During COVID-19 pandemic, doctors and Ontario Health Ministry warn against smoking marijuana — or anything | Mississauga.co

doc_id: 97da437a-f15e-47b0-b6b7-96bcff892754

the prevention and treatment of COVID-19," a Health Ministry spokesperson told the Star in an email... in the prevention and treatment of COVID-19," a Health Ministry spokesperson told the Star in an email... in the prevention and treatment of COVID-19," a Health Ministry spokesperson told the Star in an email...

FARK.com: (10791707) Smokers rejoice. Your addiction is your salvation

doc_id: 73d255ca-2b4f-494c-bb3d-f449f4b28722

So their may be some merit for marijuana as reducing effects of Covid-19.... Next up: * the Japanese saying that eating whale meat prevents Covid-19 * the Iranians saying... that eating infidels prevents Covid-19 * Trump saying that Big Macs prevent Covid-19 * Canadians saying...

EXCLUSIVE: Can smoking prevent Coronavirus infection? Experts including former FDA official point out new fac

doc_id: 3149ac1a-9b43-4441-bd6d-f1d23927827b

EXCLUSIVE: Can smoking prevent Coronavirus infection?... Experts including former FDA official point out new fact EXCLUSIVE: Can smoking prevent Coronavirus infection... When asked about the connection between smoking and COVID-19, he said: Smoking is the world's #1 preventable...

Can smoking prevent COVID-19?

Issue: can a user distinguish between correct and incorrect information?



Misinformation in the Health Domain

- People using search technologies to seek health advice online;
- COVID-19 highlighted the dangers of misinformation on consumer health;
- Uncontrolled data collections;
- Users might not be able to distinguish between correct and incorrect information;
- Incorrect documents are harmful;
- Increasing incorrect information \rightarrow users to take incorrect decisions.



Assessment Criteria

- Relevance grades: {nonrelevant, marginally relevant, fairly relevant, highly relevant};
- Relevance: multidimensional, dynamic, and complex;
- Users judge documents according to different criteria.







Agenda

How to evaluate Misinformation?

- TREC Health Misinformation;
- Evaluation with multiple aspects;
- State-of-the-art measures and their limitations;
- Problem formalisation: a partial order;
- 3 steps of TOMA;
- Example of usage;
- Experimental evaluation;
- Conclusion, limitations, and future work.

TREC Health Misinformation Track



Text Retrieval Conference (TREC)

https://trec.nist.gov/

- Large scale Evaluation Initiative (shared challenge);
- Sponsored by US NIST (National Institute of Standards and Technology);
- Started in 1992.





Health Misinformation Track

https://trec-health-misinfo.github.io/

- Track run from 2019 (Decision Track) to 2022;
- Short term goal: improve retrieval quality for health search and ability to identify misinformation;
- Long term goal: predict average decision quality of users.

TREC Health Misinformation Track

Google groups

2020 Registration

TREC Health Misinformation Track (2020)

CLARIFICATION 2020-08-4: For document identifiers (e.g., <urn:uuid:49ecaf74-b1aa-4563-83a0-c81cece0e284>) you should return only the part after the "urn:uuid:" without angle brackets (i.e., 49ecaf74-b1aa-4563-83a0-c81cece0e284).

UPDATE 2020-08-4: WET format text extracts of the corpus are now available so that you don't have to do the extraction yourself. You should use the active TREC participants password to access them.

UPDATE 2021-03-15: If you have not already registered for the task you should follow the instructions here to extract the corpus from scratch.

Track Overview



Tasks

Goal: IR systems that promote correct information over incorrect information:

- Retrieval (2019 2022): return relevant, credible and correct information (avoid returning relevant and incorrect information → harmful documents);
- Total recall (2020): identify all the documents conveying incorrect information;
- Answer prediction (2022): predict the correct answer for the topic's question.



Documents

Collection of Web Pages:

- ClueWeb12-B13 (2019): ~52 million web pages collected February 10 May 10, 2012;
- CommonCrawl News (2020): news articles from news sites collected January 1 -April 30, 2020 (focus on COVID-19);
- Noclean version of the C4 dataset (2021- 2022): ~1 billion web pages from the April 2019 snapshot of Common Crawl (used by Google to train T5).



Topics on consumer health \rightarrow people seeking health advice online;

Selection of topics: White and Hassan and Cochrane Review library.

```
<number 156</pre>
<number 156</pre>
<number 156</pre>
<query exercise scoliosis / query >
<cochranedoi > 10.1002/14051858.CD007837.pub2
</cochranedoi >
<description > Can exercises treat scoliosis?
</description >
<narrative > Scoliosis is spinal deformity,
which occurs as sideways curvature, ... A
relevant document discusses whether
exercises can help to treat scoliosis or
improve lives of people with scoliosis.
// narrative >
</topic >
```



Focus on COVID 19 → treatment is alway COVID-19

Selection of topics: WHO mythbusters, Harvard Medical School and Fact checking websites (Snopes, etc.)



Selection of topics: authoritative sources;

```
<topic>
<number>1234</number>
<query>dexamethasene croupk/query>
 description Is dexamethasone a good
treatment for croup?</description>
<narrative>Croup is an infection of the
upper airway ... A useful document would
provide information that would help a user
make a decision about treating croup with
dexamethasone, and may discuss either
separately or jointly: croup, recommended
treatments for croup, the pros and cons of
dexamethasone, etc.</narrative>
<disclaimer>We do not claim to be providing
medical advice, and medical decisions
should never be made based on the stance we
have chosen. Consult a medical doctor for
professional advice.</disclaimer>
<stance>helpful</stance>
<evidence>https://www.ncbi.nlm.nih.gov/pmc/
articles/PMC5804741/</evidence>
</topic>
```

Treatment Disease

No more structured sentence (yes/no answer)



Selection of topics: from real queries submitted to a search engine;

```
<topic>
<number>12345</number>
<question>Does apple cider vinegar work to
treat ear infections?</guestion>
<query>apple cider vinegal ear infection
query>
<background>Apple cider vinegar is a common
cooking ingredient that contains...
bacteria and cause fluid build up in the
middle ear, which is located behind the
eardrum.</background>
<disclaimer>We do not claim to be providing
medical advice, and medical decisions
should never be made based on the answer we
have chosen. Consult a medical doctor for
professional advice.</disclaimer>
</topic>
```



Answer and evidence were provided after the submission (answer prediction task).



Judgements

- Assessors did not create the topics;
- Assessors did not have access to the topic answer;
- All aspects are assessed independently;
- Assessed aspects:
 - Relevance, Treatment Efficacy, Credibility (2019);
 - Usefulness, Answer, Credibility (2020);
 - Usefulness, Supportiveness, Credibility (2021);
 - Usefulness, Answer + Preference Judgements (2022).

Judging Credibility

- Google Search Quality Evaluator Guidelines:
 - Understand the **purpose** of a document;
 - Amount of Expertise, Authoritativeness, and Trustworthiness (E-A-T).
- Credible document:
 - High level of E-A-T;
 - Includes an author or a publishing institute expert in the field;
 - Includes citations or references to credible sources, e.g., universities, research/clinics, government websites, etc.;
 - Hosted in a hospital/clinic or government website, or online newspaper with wide circulation;
 - Style: well written, motivated and organized.
- Not credible document:
 - Advertising or marketing purposes, from a personal blog or a forum, or written by a non-expert person;
 - Document or the hosting website provides or claims against well-known medical consensus (e.g., smoking cigarettes does not cause cancer).

Assessments in 2019Relevance, Treatment Efficacy, Credibility

- Topic:
 - Helpful: The health treatment is helpful towards the health issue.
 - Inconclusive: It is still unclear by medical professionals whether or not the treatment is effective towards the health issue.
 - Not helpful: The treatment is not helpful towards the health issue.
- Assessed aspects:
 - Relevance (3 levels): Highly relevant, relevant, not relevant;
 - Efficacy (4 values): Effective, inconclusive, ineffective, no information;
 - Credibility (binary): Credible, not credible.
- Efficacy and credibility → only if the document is relevant;
- Correctness: derived by comparing topic efficacy against document efficacy.

Assessments in 2020

- Topic answer: "yes" or "no" → removed inconclusive topics;
- Usefulness (binary): does the document contain material that the user might find useful in answering the topic's question?
- Answer (3 values: yes, no, no answer): does the document answer to the question in the description field? If so, is the answer yes or no?
- Credibility (binary): how credible is the document?
- Answer and credibility only for useful documents;
- Correctness: derived by comparing topic answer with assessor answer → incorrect, correct, no answer.

Assessments in 2021

- Topic answer: "helpful" or "unhelpful";
- Usefulness (3 levels): does the document contain material that the user might find useful in answering the topic's question?
- Supportiveness (3 values: supports, dissuades, neutral): does the document contain information that supports/dissuades the use of the treatment in the question?
- Credibility (3 levels): how credible is the document?
- Supportiveness and credibility only for useful documents;
- Correctness: derived by comparing topic answer with assessor answer → incorrect, correct, no answer.

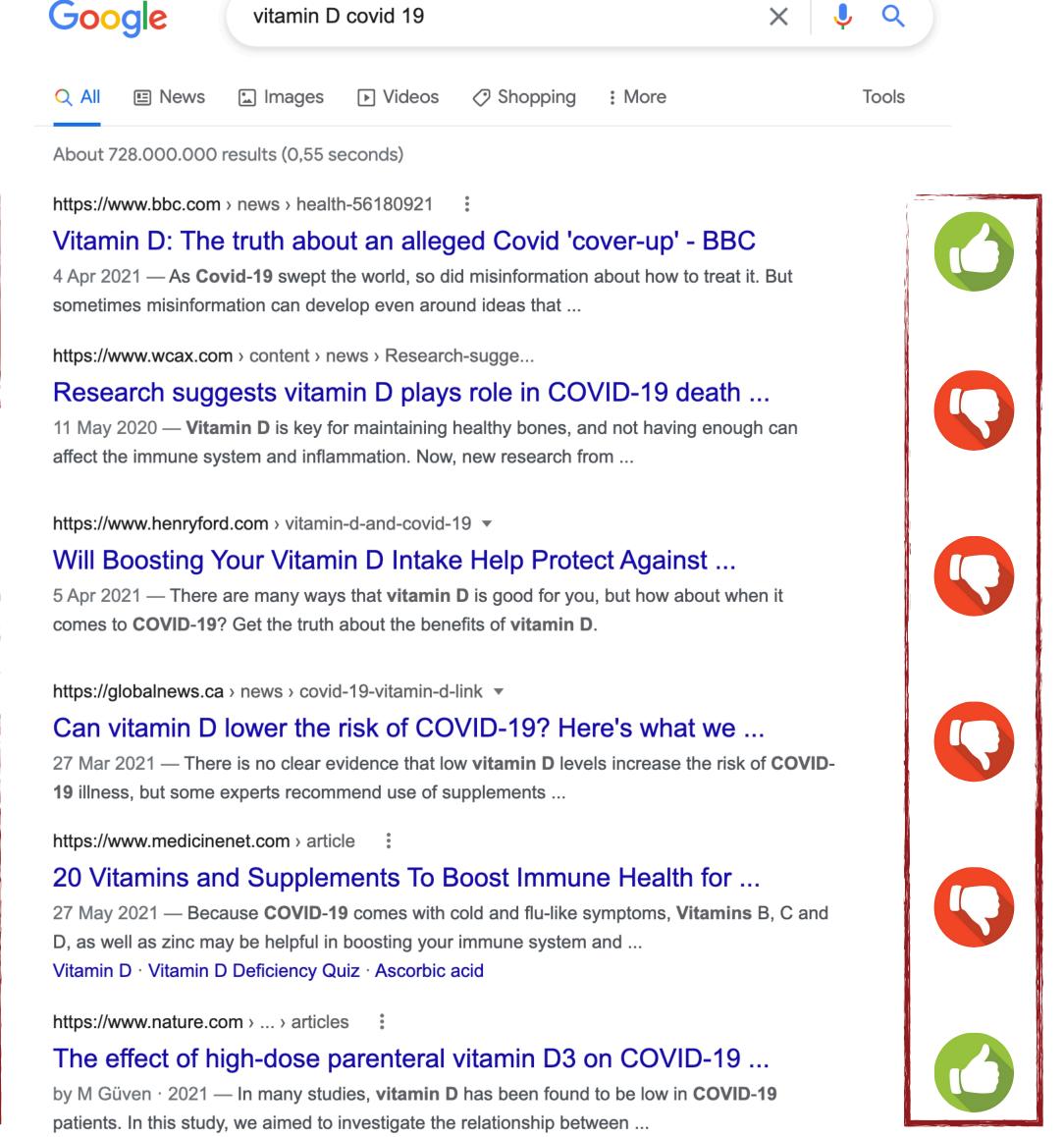
Assessments in 2022

- Topic answer: "yes" or "no";
- Usefulness (3 levels): does the document contain material that the user might find useful in answering the topic's question?
- Answer (3 values: yes, no, no answer): does the document answer to the question in the description field? If so, is the answer yes or no?
- Correctness: derived by comparing topic answer with assessor answer → incorrect, correct, no answer.
- Preference judgements: top 10 very useful documents per topic.

Multi-aspect Evaluation

Relevance Labels

Multi-aspect Evaluation



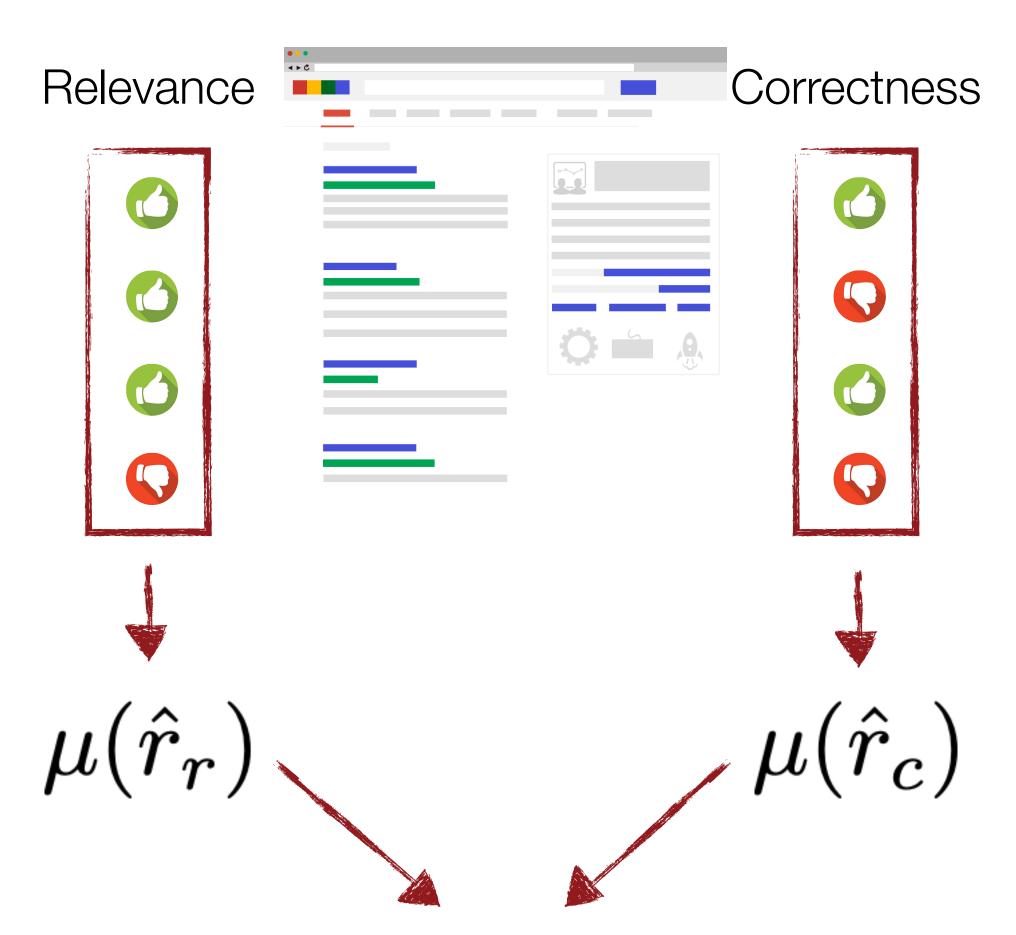




How to account for multiple aspects in IR evaluation?



State-of-the-art Measures: Approach 1



Measures that apply this approach:

- Convex Aggregating Measure (CAM);
- Multidimensional Measure (MM);
- Weighted Harmonic Mean Aggregating Measure (WHAM).

J. Palotti, G. Zuccon, and A. Hanbury. 2018. MM: A New Framework for Multidimensional Evaluation of Search Engines. In CIKM 2018.

C. Lioma, J. G. Simonsen, and B. Larsen. 2017. Evaluation Measures for Relevance and Credibility in Ranked Lists. In ICTIR 2017.

Aggregation $(\mu(\hat{r}_r), \mu(\hat{r}_c))$

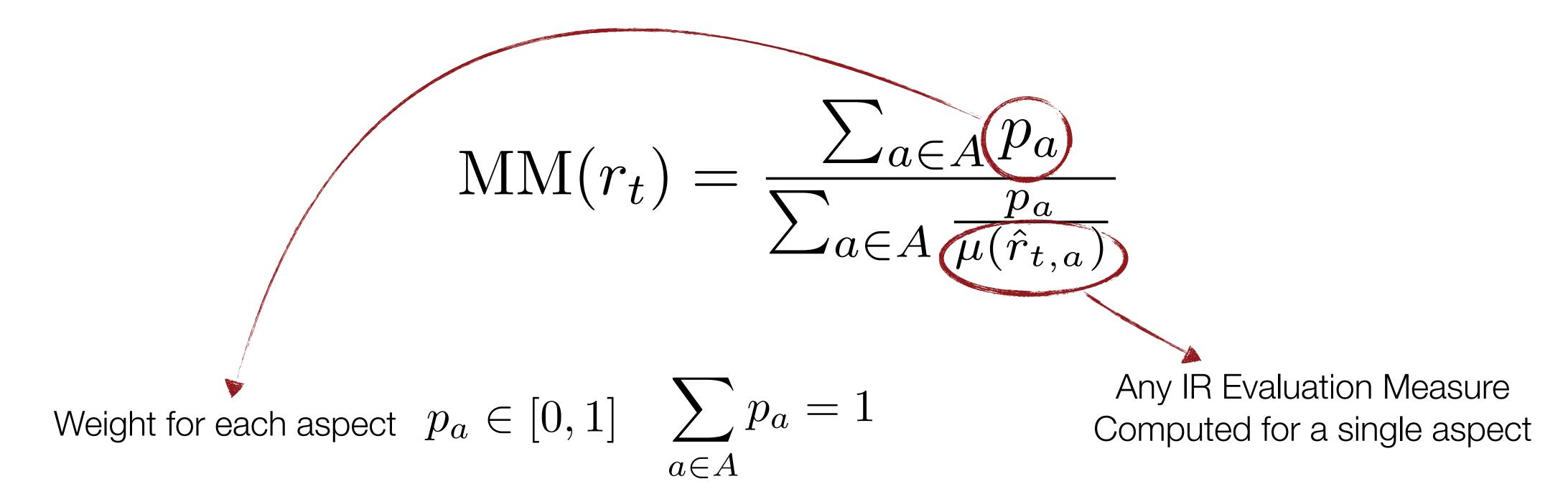
Convex Aggregating Measure (CAM)

Mean over measure scores across aspects:

$$\mathrm{CAM}(r_t) = \sum_{a \in A} p_a \times \mu(\hat{r}_{t,a})$$
 Weight for each aspect $p_a \in [0,1]$
$$\sum_{a \in A} p_a = 1$$
 Any IR Evaluation Measure Computed for a single aspect

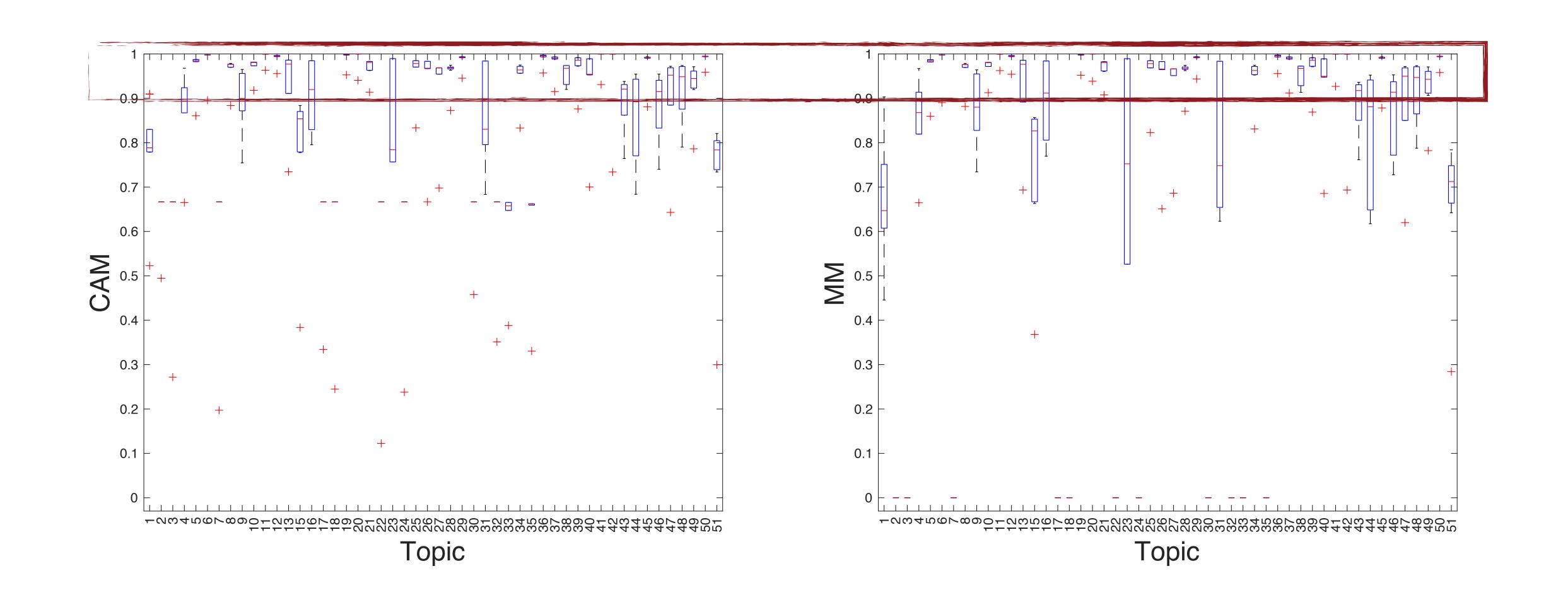
Multidimensional Measure (MM)

Harmonic mean over measure scores across aspects:



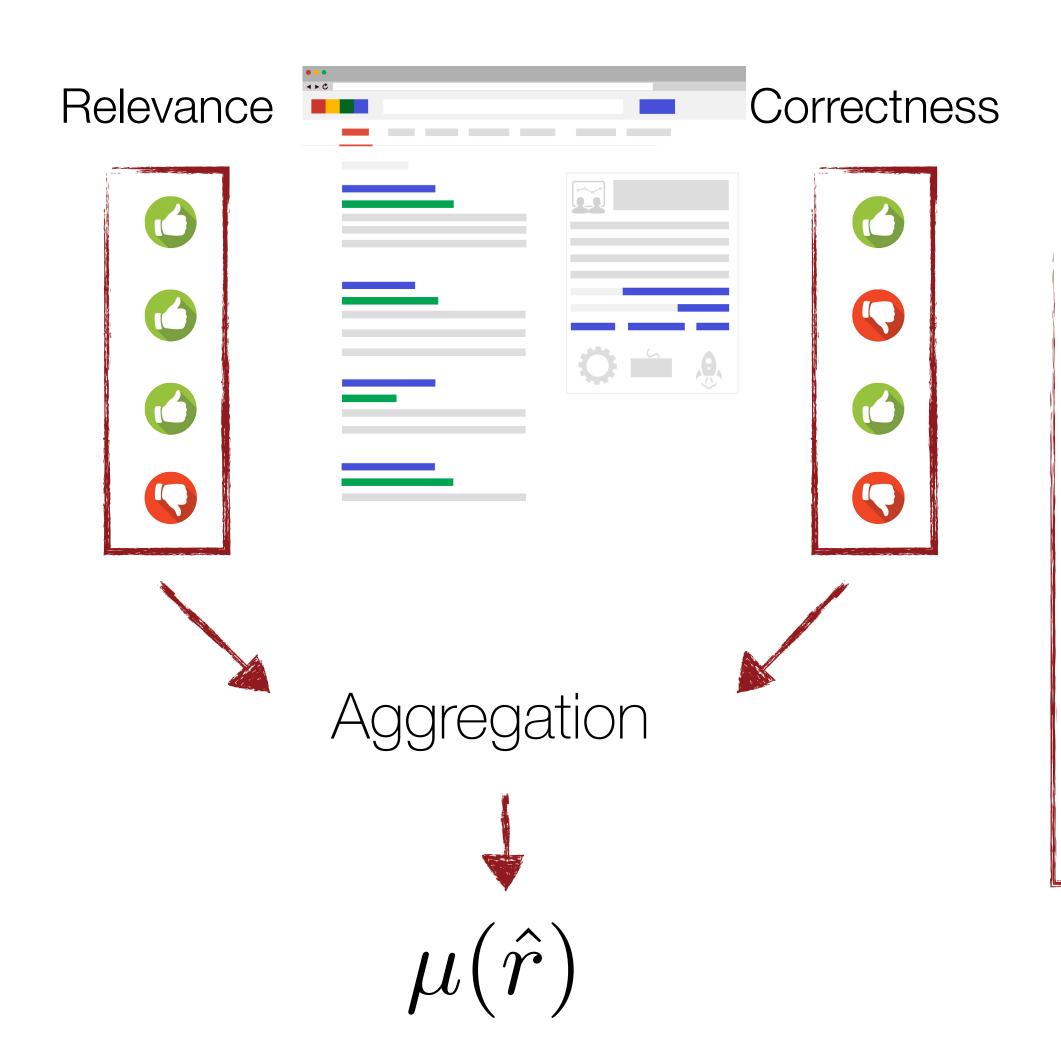


Limitations of Approach 1





State-of-the-art Measures: Approach 2



Measures that simultaneously evaluates with all aspects:

- α-nDCG: relevance and diversity;
- nCT: relevance, novelty and user effort;
- RBU: relevance, redundancy and user effort;
- nLRE: relevance and credibility;
- etc.

E. Amigó, D. Spina, and J. Carrillo-de Albornoz. 2018. An Axiomatic Analysis of Diversity Evaluation Metrics: Introducing the Rank-Biased Utility Metric. In SIGIR 2018.

C. L. A. Clarke, M. Kolla, G. V. Cormack, O. Vechtomova, A. Ashkan, S.Büttcher, and I. MacKinnon. 2008. Novelty and Diversity in Information Retrieval Evaluation. In SIGIR 2008.

Z. Tang and G. H. Yang. 2017. Investigating per Topic Upper Bound for Session Search Evaluation. In ICTIR '17.

Normalized Local Rank Error (NLRE)

- Error computed with respect to the ideal ranking for each aspect separately;
- Error = difference in rank positions;
- Computes the maximum possible error attainable:

LRE =
$$\sum_{i=1}^{n-1} \frac{1}{\log_2(1+i)} \left(\prod_{a \in A} (p_a) + \epsilon_a[d_i] \right) - \prod_{a \in A} p_a \right) \qquad \text{NLRE} = 1 - \frac{\text{LRE}}{C_{\text{LRE}}}$$

$$NLRE = 1 - \frac{LRE}{C_{LRE}}$$

Weight for each aspect $p_a \in [0,1]$

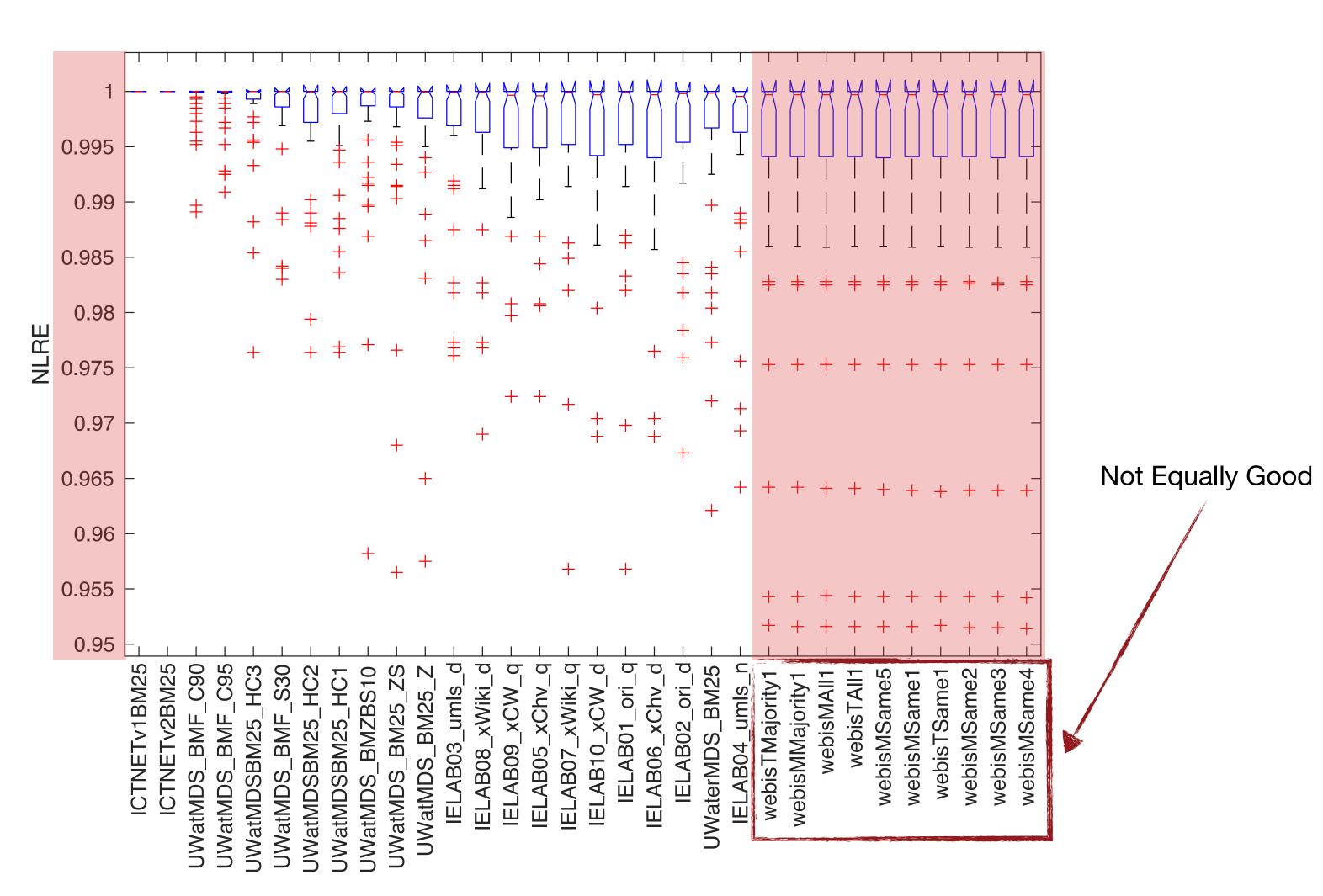
Normalization factor: all possible permutations



Limitations of Approach 2

Evaluation Measure Score

- Most measures are defined for specific contexts and with a limited set of aspects;
- Most measures do not define a formal framework, i.e., the ideal ranking is not defined or it is hard to compute.





Compatibility

- Define a preference order (and an ideal ranking);
- Rank Biased Overlap (RBO);
- Compute RBO between the run and the ideal ranking;
- For helpful and harmful documents independently.

Preference Value	Usefulness	Correctness	Credibility
12	Very Useful	Correct	Excellent
11	Useful	Correct	Excellent
10	Very Useful	Correct	Good
9	Useful	Correct	Good
8	Very Useful	Correct	Low or Not Judged
7	Useful	Correct	Low or Not Judged
6	Very Useful	Neutral or Not Judged	Excellent
5	Useful	Neutral or Not Judged	Excellent
4	Very Useful	Neutral or Not Judged	Good
3	Useful	Neutral or Not Judged	Good
2	Very Useful	Neutral or Not Judged	Low or Not Judged
1	Useful	Neutral or Not Judged	Low or Not Judged
0	Not Useful	Not Judged	Not Judged
-1	Very Useful or Useful	Incorrect	Low or Not Judged
-2	Very Useful or Useful	Incorrect	Good
3	Very Useful or Useful	Incorrect	Excellent

Problem Formalization: Partial Order



Why a Formal Framework?

- Two aspects, relevance and correctness;
- Relevance Labels: {highly relevant, fairly relevant, marginally relevant, not relevant};
- Correctness Labels: {correct, partially correct, not correct};

```
d_1 (highly relevant, correct) d_2 \sqsubseteq d_1 d_3 \sqsubseteq d_1 d_2 ? d_3 (highly relevant, partially correct)
```



Formalization of the Problem

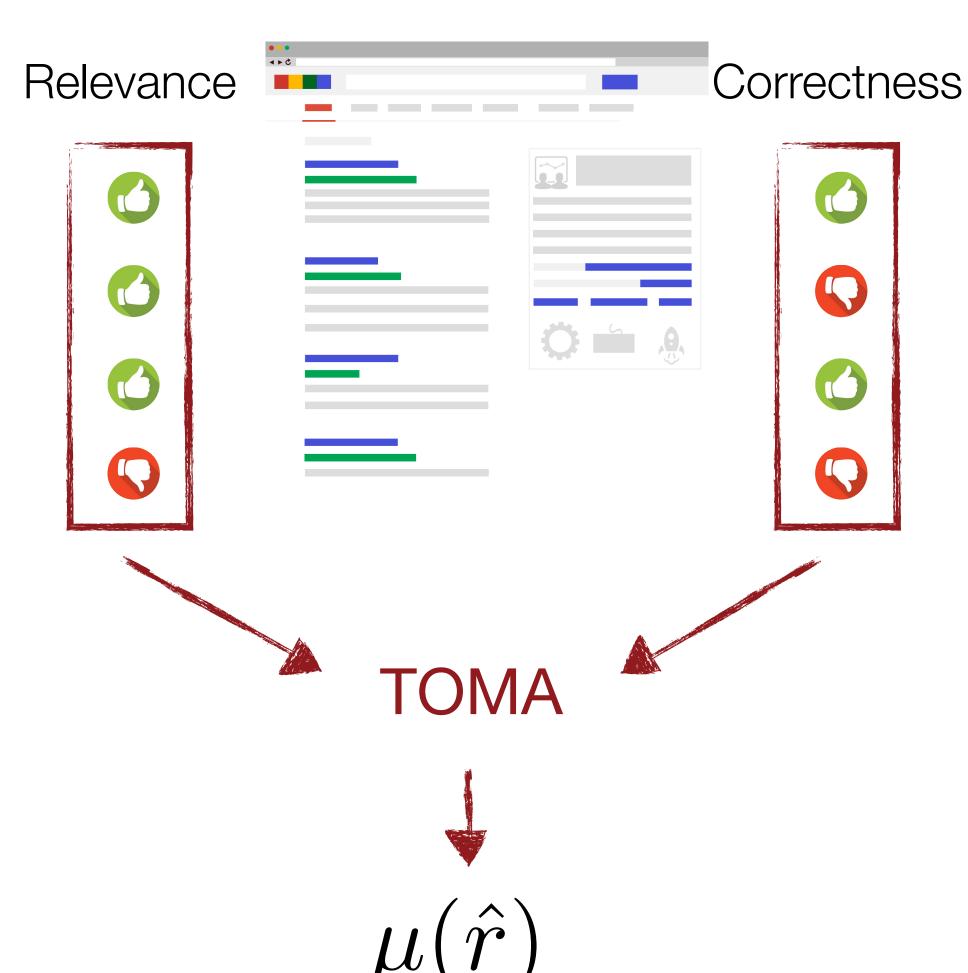
- n aspects $A = \{a_1, a_2, \ldots, a_n\}$, for example $A = \{\text{relevance, correctness}\}$
- Set of labels $l_0^a \prec_a l_1^a \prec_a \ldots \prec_a l_{K_a}^a$, for example non-rel $< \ldots <$ highly rel
- Ground-truth function: $GT(d,t) = (l_1, l_2, \ldots, l_n)$
- Unequivocal order: $GT(d,t) \sqsubseteq GT(d',t) \iff l_i \preceq_{a_i} l'_i \quad \forall i \in \{1,\ldots,n\}$
- Partial order, i.e., there are elements that are not comparable
- Non comparable documents: $GT(d,t) \not\sqsubseteq GT(d',t)$ $GT(d',t) \not\sqsubseteq GT(d,t)$

 d_2 (marginally relevant, correct) d_3 (highly relevant, partially correct)

TOMA Evaluation



How to Complete the Partial Order?



Total Order Multi-Aspect (TOMA) evaluation:

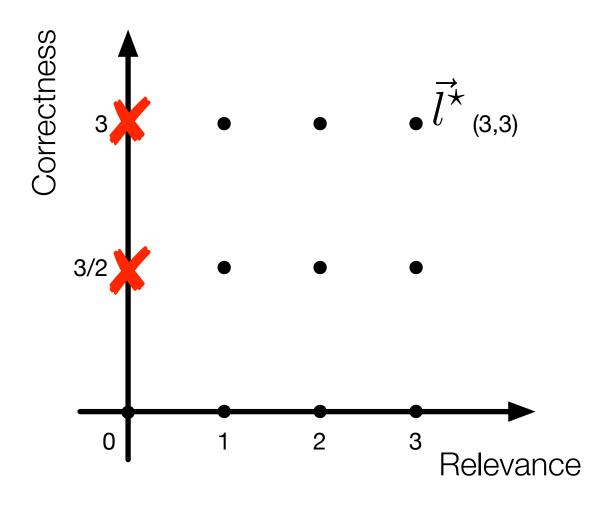
- Step 1: embedding in the Euclidean space;
- Step 2: the distance function & the distance order;
- Step 3: the weight function;

Compute any IR measure μ .



TOMA Example

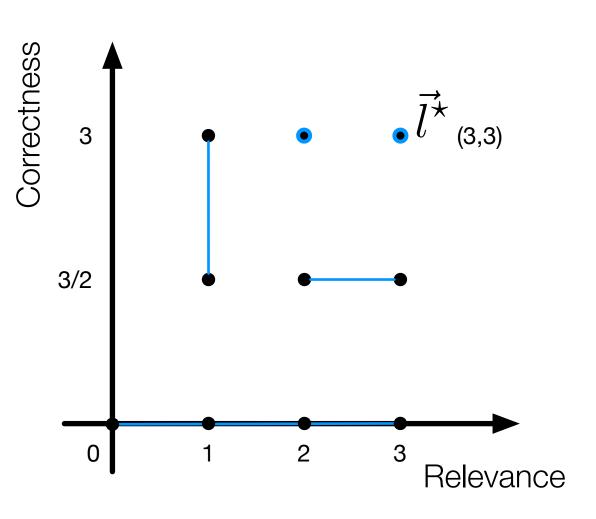
Step 1: Embedding



Relevance = {not relevant < marginally relevant < fairly relevant < highly relevant}

Correctness = {not correct <
partially correct < correct}</pre>

Step 2: Distance Function



$$Dist(\vec{l}, \vec{l}^{\star}) = \max\{|g_r(l^{\star}) - g_r(l)|, |g_c(l^{\star}) - g_c(l)|\}$$

Step 3: Weight Function

$$[(3,3)] = \{(3,3)\} \qquad \mapsto 4$$

$$[(2,3)] = \{(2,3)\} \qquad \mapsto 3$$

$$[(3,3/2)] = \{(3,3/2), (2,3/2)\} \qquad \mapsto 2$$

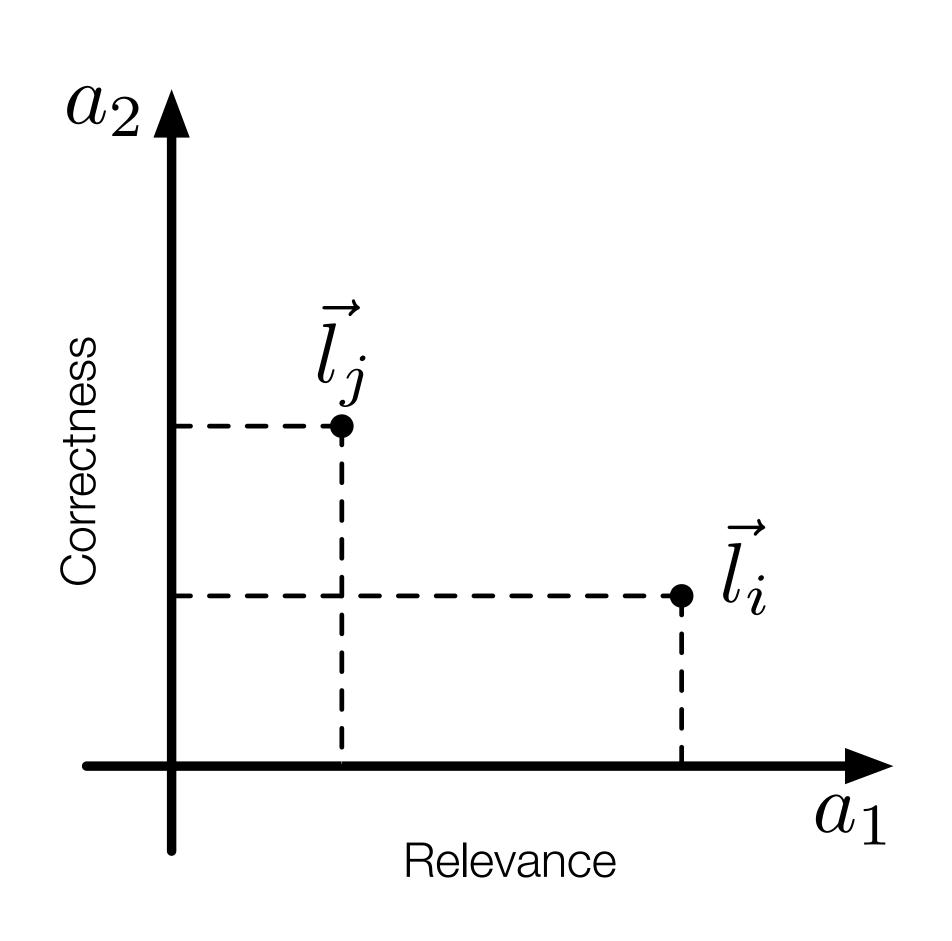
$$[(1,3)] = \{(1,3), (1,3/2)\} \qquad \mapsto 1$$

$$[(0,0)] = \{(0,0), (1,0), (2,0), (3,0)\} \qquad \mapsto 0$$



TOMA Step 1

Embedding in the Euclidean space and Distance Order



• Embedding function g:

$$\vec{l} = g(l_1, \dots, l_n) = (g_{a_1}(l_1), \dots, g_{a_n}(l_n))$$

- Best label tuple (maximum element) l^* ;
- Compute the distance from the best label:

$$\mathbf{l} \preceq_* \mathbf{l}' \iff \mathrm{Dist}(\vec{l}, \vec{l^{\star}}) \geq \mathrm{Dist}(\vec{l'}, \vec{l^{\star}})$$

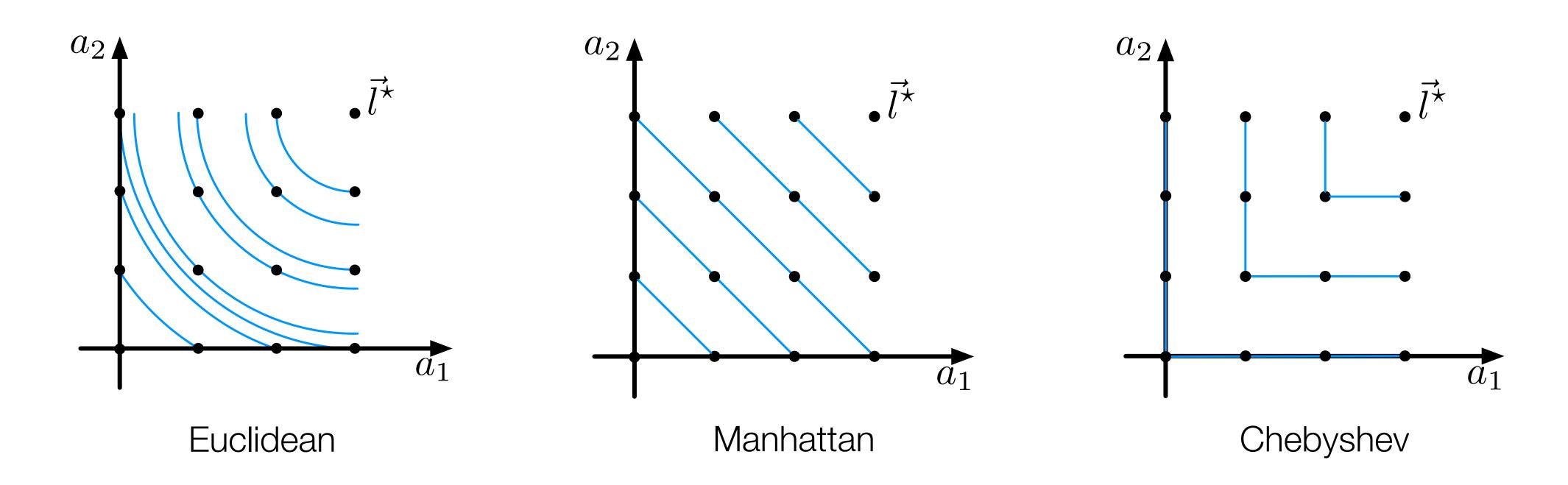
• Distance order \leq_* , a weak order relation:

$$\mathbf{l} =_* \mathbf{l}' \iff \operatorname{Dist}(\vec{l}, \vec{l}^*) = \operatorname{Dist}(\vec{l'}, \vec{l'})$$



TOMA Step 2

The Distance Function



• The distance order is a weak order, so we can define equivalence classes:

$$[\mathbf{l}]_{\star} = \{\mathbf{l}' \in L : \operatorname{Dist}(\vec{l'}, \vec{l^{\star}}) = \operatorname{Dist}(\vec{l}, \vec{l^{\star}})\}$$



TOMA Step 3

The Weight Function

- The Weight Function $W:L\to\mathbb{N}_0^+$, maps each equivalence class to a weight (non-negative integer);
- The distance order relation must be preserved:

$$\forall \mathbf{l}, \mathbf{l}' \in L : \mathbf{l} \preceq_* \mathbf{l}' \Longrightarrow W(\mathbf{l}) \leq W(\mathbf{l}')$$

• Once the weight function is defined, we can compute any IR measures on the multi-aspect ranking r_t :

$$\mu \circ W \colon M(r_t) = \mu \Big(W(GT(d_1, t)), \dots, W(GT(d_N, t)) \Big)$$



Partial Order vs. Distance Order

We started with a partial order:

$$GT(d,t) \sqsubseteq GT(d',t) \iff$$

$$l_i \preceq_{a_i} l'_i \quad \forall i \in \{1,\ldots,n\}$$

- TOMA respects this partial order;
- $\forall 1, 1' \in L$ we have:

$$1 \sqsubset 1' \implies 1 \prec_* 1'$$

Appendix of Principled Multi-Aspect Evaluation of Rankings

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A.1 Proof 1: TOMA Respects the Partial Order

Given a set of documents D, a set of aspects A, a set of topics T, and a ground-truth map GT a partial order on the set of tuples of labels $L = X_{a \in A} L_a$ is defined as follows:

$$GT(d,t) \sqsubseteq GT(d',t) \iff l_i \leq_{a_i} l'_i \quad \forall i \in \{1,\ldots,n\}$$
 (1)

where $GT(d, t) = (l_1, ..., l_n)$ and $GT(d', t) = (l'_1, ..., l'_n)$.

Let g be an embedding function that maps tuples of labels in Euclidean space $\mathcal{L} = \mathbb{R}^n$: $g(\mathbf{l}) = g(l_1, \ldots, l_n) = (g_{a_1}(l_1), \ldots, g_{a_n}(l_n))$. We assume that for each $a \in A$, g_a is a non-decreasing map, i.e., for any $l, l' \in L_a$ if $l \leq_a l'$ than $g_a(l) \leq g_a(l')$. Through the embedding function g, each tuple of labels \mathbf{l} is represented by a point in the Euclidean space \mathcal{L} denoted by $\vec{l} = g(\mathbf{l})$.

We define the *distance order* \leq_* : a weak order on L such that:

$$1 \leq_* 1' \iff \operatorname{Dist}(\vec{l}, \vec{l^*}) \geq \operatorname{Dist}(\vec{l'}, \vec{l^*})$$
 (2)

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and by considering the absolute value:

$$|g_a(l_a) - g_a(l_a^*)| \ge |g_a(l_a') - g_a(l_a^*)| \quad \forall a \in A$$

which means that $\operatorname{Dist}(\vec{l}, \vec{l}^*) \geq \operatorname{Dist}(\vec{l'}, \vec{l'})$ with Manhattan distance, i.e. $1 \leq_1 1'$.

Analogously, taking the square values in Equation (7), we obtain:

$$\left(g_a(l_a) - g_a(l_a^{\star})\right)^2 \ge \left(g_a(l_a') - g_a(l_a^{\star})\right)^2 \quad \forall a \in A \tag{9}$$

which implies that $\operatorname{Dist}(\vec{l}, \vec{l}^{\star}) \geq \operatorname{Dist}(\vec{l}', \vec{l}^{\star})$ with Euclidean distance, i.e. $1 \leq_2 1'$.

To conclude, since $|A| < \infty$ there exists $\bar{a} \in A$ such that:

$$\max_{a \in A} |g_a(l_a') - g_a(l_a^*)| = |g_{\bar{a}}(l_{\bar{a}}') - g_{\bar{a}}(l_{\bar{a}}^*)| \tag{10}$$

by Equation (9) we have:

$$|g_{\bar{a}}(l_{\bar{a}}') - g_{\bar{a}}(l_{\bar{a}}^{\star})| \le |g_a(l_a) - g_a(l_a^{\star})| \le \max_{a \in A} |g_a(l_a) - g_a(l_a^{\star})|$$
 (11)

 $\forall a \in A$ which implies that $\text{Dist}(\vec{1} \mid \vec{1} \star) \times \text{Dist}(\vec{1} \mid \vec{1} \star)$ with Chahr



Example of Usage

- Two aspects: relevance and correctness;
- Assumption: not relevant documents are not correct;
- Three different embedding functions.

Relevance	Correctness	Distance	Order among Tuples of Labels
{0, 1, 2, 3}	{0, 3/2, 3}	Euclidean Manhattan Chebyshev	$(3,3) \leq_* (2,3) \leq_* (3,3/2) \leq_* (2,3/2) \leq_* (1,3) \leq_* (1,3/2) \leq_* (3,0) \leq_* (2,0) \leq_* (1,0) \leq_* (0,0)$ $(3,3) \leq_* (2,3) \leq_* (3,3/2) \leq_* (1,3) \leq_* (2,3/2) \leq_* (3,0) \leq_* (1,3/2) \leq_* (2,0) \leq_* (1,0) \leq_* (0,0)$ $(3,3) \leq_* (2,3) \leq_* (3,3/2) =_* (2,3/2) \leq_* (1,3) =_* (1,3/2) \leq_* (3,0) =_* (2,0) =_* (1,0) =_* (0,0)$
{0, 1, 2, 3}	$\{0, 1, 2\}$	Euclidean Manhattan Chebyshev	$(3,2) \leq_* (3,1) =_* (2,2) \leq_* (2,1) \leq_* (3,0) =_* (1,2) \leq_* (2,0) =_* (1,1) \leq_* (1,0) \leq_* (0,0)$ $(3,2) \leq_* (3,1) =_* (2,2) \leq_* (3,0) =_* (2,1) =_* (1,2) \leq_* (2,0) =_* (1,1) \leq_* (1,0) \leq_* (0,0)$ $(3,2) \leq_* (3,1) =_* (2,1) =_* (2,2) \leq_* (3,0) =_* (2,0) =_* (1,0) =_* (1,1) =_* (1,2) \leq_* (0,0)$
{0, 1, 2, 3}	{0, 2, 6}	Euclidean Manhattan Chebyshev	$(3,6) \leq_* (2,6) \leq_* (1,6) \leq_* (3,2) \leq_* (2,2) \leq_* (1,2) \leq_* (3,0) \leq_* (2,0) \leq_* (1,0) \leq_* (0,0)$ $(3,6) \leq_* (2,6) \leq_* (1,6) \leq_* (3,2) \leq_* (2,2) \leq_* (1,2) =_* (3,0) \leq_* (2,0) \leq_* (1,0) \leq_* (0,0)$ $(3,6) \leq_* (2,6) \leq_* (1,6) \leq_* (3,2) =_* (2,2) =_* (1,2) \leq_* (3,0) =_* (2,0) =_* (1,0) =_* (0,0)$

Experiments



Experimental Set-up

Datasets

	TREC tracks									
	Web 2009	Web 2010	Web 2011	Web 2012	Web 2013	Web 201	4 Task 2015	Task 2016	Decision 2019	Misinfo2020
Collection		ClueWe	b09			Clı	ıeWeb12	ClueWeb12-B13	CommonCrawl News	
Topics	50	48	50	50	50	50	35	50	50	46
Submitted runs	71	56	61	48	61	30	6	9	32	51
Aspects (label grades)	relevance (4) popularity† (3) non-spam‡ (3)	relevance (5*) popularity† (3) non-spam‡ (3)	relevance (4*) popularity† (3) non-spam‡ (3)	1	relevance (5* popularity† (3 non-spam‡ (3	3)	usefu popula	ance (3*) lness (3) arity† (3) pam‡ (3)	relevance (3) credibility (2) correctness (2)	relevance (2) credibility (2) correctness (2)

- 10 TREC Tracks, 425 runs;
- Up to 5 different aspects;
- Popularity estimated with PageRank Score¹;
- Non-spamminess estimated with Waterloo Spam Ranking².

^[1] http://www.lemurproject.org/clueweb12/PageRank.php

^[2] https://www.mansci.uwaterloo.ca/~msmucker/cw12spam/



Experimental Set-up

Baselines and TOMA Versions

We use CAM and MM as baselines:

$$CAM(r_t) = \sum_{a \in A} p_a \times \mu(\hat{r}_{t,a})$$

$$MM(r_t) = \frac{\sum_{a \in A} p_a}{\sum_{a \in A} \frac{p_a}{\mu(\hat{r}_{t,a})}}$$

where $p_a \in [0,1]$ and $\sum_{a \in A} p_a = 1$;

- 3 Versions of TOMA: EUCL (Euclidean), MANH (Manhattan), CHEB (Chebyshev);
- IR Measures: nDCG (graded labels when available) and AP (binary labels);
- All aspects are embedded into integer values with unit step 1.



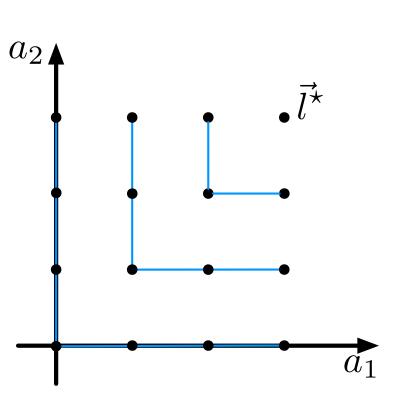
Correlation Analysis

- Kendall's τ between the Rankings of Submitted runs (RoS) generated by 2 different measures;
- Most correlated RoS:
 - EUCL MANH (τ up to 1);
 - (EUCL, MANH) CAM (τ up to 0.76);
 - CHEB MM (τ up to 0.88).
- EUCL and MANH may be more similar to CAM (mean), while CHEB may be more similar to MM (harmonic mean).



Discriminative Power

- Bootstrap Samples: 10'000
- Paired Bootstrap Hypothesis Test, with $\alpha = 0.01$;



• 16/20 times either MANH (12/20) or EUCL (6/20) is best.

	WEB	2009	WEB	2010	WEB2	2011	WEB:	2012	WEB:	2013	WEB	2014	TAS	K15	TAS	K16	DECISI	ON19	MISINF	O 2020
	NDCG	AP	NDCG	AP	NDCG	AP	NDCG	AP	NDCG	AP	NDCG	AP	NDCG	AP	NDCG	AP	NDCG	AP	NDCG	AP
	İ							DIS	CRIMINA	TIVE PC	WER OF	MEASU	RES							
CAM	75.98	64.43	66.32	61.23	75.14	61.64	68.71	56.74	76.89	57.05	85.06	78.85	53.33	33.33	72.22	55.56	72.58	70.56	71.53	70.90
MM	75.61	50.58	72.89	67.79	67.32	67.81	62.68	56.12	80.71	46.99	74.25	53.56	0.00	0.00	0.00	0.00	60.08	53.23	68.31	62.20
EUCL	75.29	72.64	62.96	66.75	75.14	70.33	66.13	64.10	75.14	59.45	80.92	78.85	66.67	66.67	69.44	75.00	73.59	73.99	72.86	75.14
MANH	76.66	72.68	63.59	67.14	77.32	70.38	66.05	64.18	76.67	59.34	86.44	79.08	66.67	53.33	75.00	75.00	73.79	73.79	73.02	74.98
CHEB	50.18	6.32	59.82	51.49	73.06	50.11	61.08	39.36	77.10	49.34	75.17	66.21	0.00	0.00	0.00	0.00	42.54	29.84	65.41	53.33



Document Quality

Run that has been assessed as the best best per {topic, track, year}, on a per query basis;

- With cut-off 5, zero-Aspect: # documents with 0 as sum of labels;
- Average per label: average across aspects.

Zero-Aspect Documents

Rank	CAM	MM	EUCL	MANH	CHEB
1	51 (1.18%)	131 (3.02%)	39 (0.90%)	33 (0.76%)	154 (3.55%)
2	65 (1.50%)	159 (3.67%)	50 (1.15%)	48 (1.11%)	179 (4.13%)
3	103 (2.38%)	202 (4.66%)	88 (2.03%)	78 (1.80%)	185 (4.17%)
4	102 (2.35%)	173 (3.99%)	86 (1.99%)	74 (1.71%)	183 (4.23%)
5	107 (2.47%)	196 (4.53%)	95 (2.19%)	81 (1.87%)	205 (4.73%)
1-5	428 (9.88%)	861 (19.88%)	358 (8.27%)	314 (7.25%)	906 (20.92%)

Average per Label

Ranks	CAM	MM	EUCL	MANH	CHEB
1-25	1.70	1.49	1.67	1.69	1.39
26-50	0.85	0.78	0.91	0.94	0.70
51-75	0.57	0.53	0.63	0.64	0.48
76-100	0.40	0.39	0.43	0.44	0.36



Limitations

- Arbitrary choices of the embedding, distance, and weight functions → impact of each choice;
- Embedding function: mapping from nominal/ordinal to interval/ratio scale → theoretical properties of TOMA;
- Analysis of the interactions between aspects and/or documents → proper embedding;
- Large number of equivalence classes might be a problem for gain based measures → non-gain based measures.



Wrapping up

Total Order Multi-Aspect (TOMA):

- Defined for any number and type of aspect;
- Aspects can have different gradings;
- Choose a relative importance weighting for different aspects;
- Integration with any existing single-aspect IR measure;
- Better discriminative power;
- Better at rewarding high quality documents.

Conclusions and Future Work



In Today's Talk

- TREC Health Misinformation track;
- Challenges in evaluating misinformation: find good topics, collect judgements, datasets with enough noise, etc.
- Multi-aspect evaluation and existing measures;
- Total Order Multi-Aspect (TOMA): defined for any number and type of aspect with different gradings integrated with any existing single-aspect IR measure.

Future Work

- TREC Trustworthy Al Conference (TRUC);
- Formal properties of evaluation measures;
- Measures that penalise systems that retrieve harmful documents;
- Other aspects, e.g., fairness;
- Use these measures for learning to rank;
- Relation to with bias;
- Ethical issues.

Thank You!

Questions?



Ongoing Work



Relation with Other Measures - uRBP

Understandability Biased RBP (uRBP):

$$\label{eq:urbance} \begin{split} \text{uRBP@}k(r_t) &= (1-p)\sum_{i=1}^k p^{i-1} \hat{r}_{t,r}[i] \cdot \hat{r}_{t,u}[i] \end{split}$$
 Relevance Understandability

Within TOMA framework:

$$[\mathbf{l}] = \left\{ \mathbf{l}' \in L : \prod_{a \in A} g_a(\mathbf{l}') = \prod_{a \in A} g_a(\mathbf{l}) \right\}$$

Relation with Other Measure - Compatibility

- Compute Compatibility:
 - 1. Define an ideal ranking or set of equivalent ideal rankings;
 - 2. Compute RBO between the input ranking and ideal ranking;

RBO
$$(r_t, i_t) = (1 - p) \sum_{i=1}^{+\infty} p^{i-1} \cdot \frac{|r_t[1, i] \cap i_t[1, i]|}{i}$$

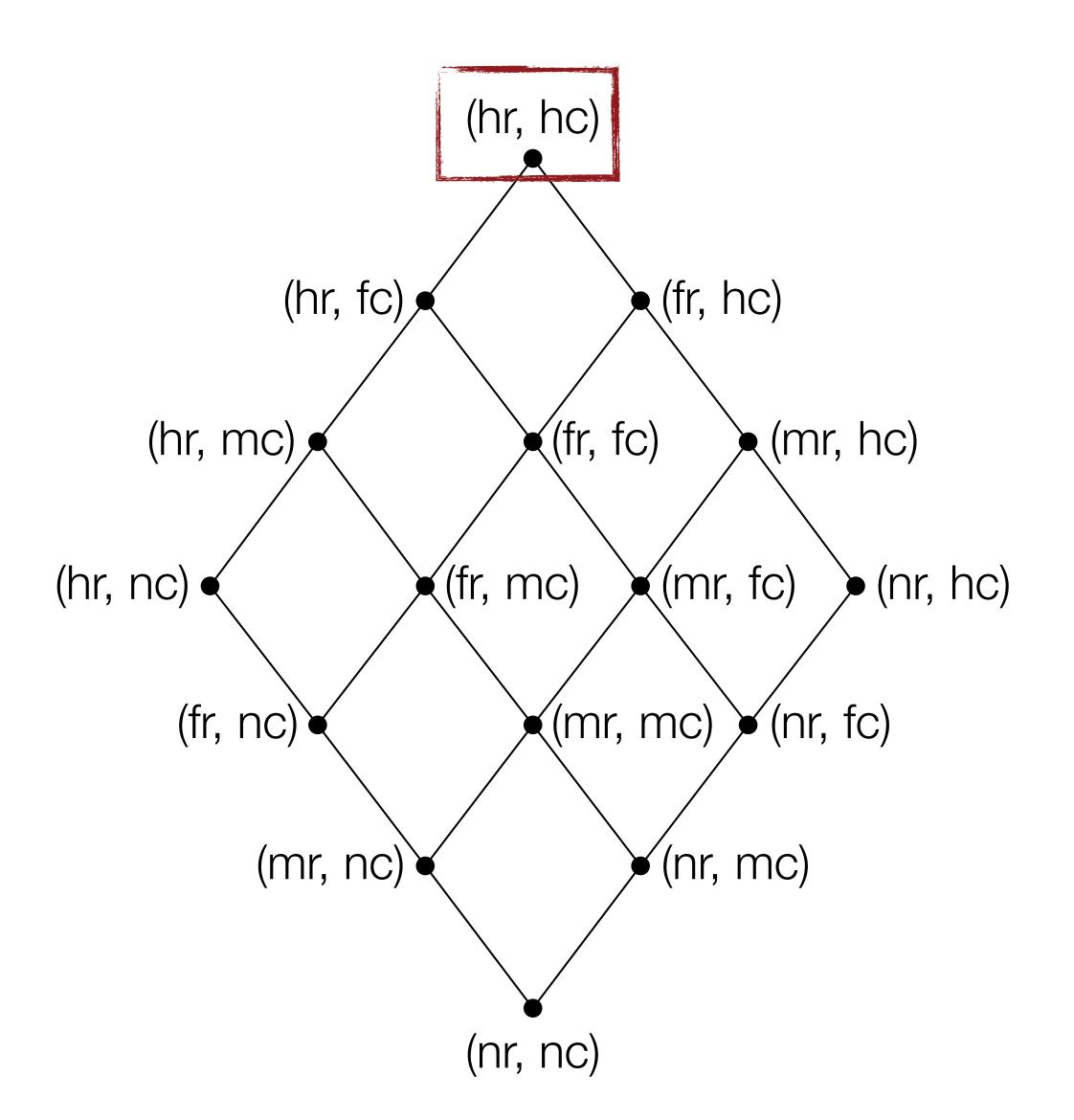
Use TOMA to define levels of effectiveness ~ equivalence classes.



Poset Theory

- Poset: partially ordered set;
- Set of multi-aspect labels is a poset.

hr highly relevant hc highly correct
fr fairly relevant fc fairly correct
mr marginally relevant mc marginally correct
nr not relevant nc not correct



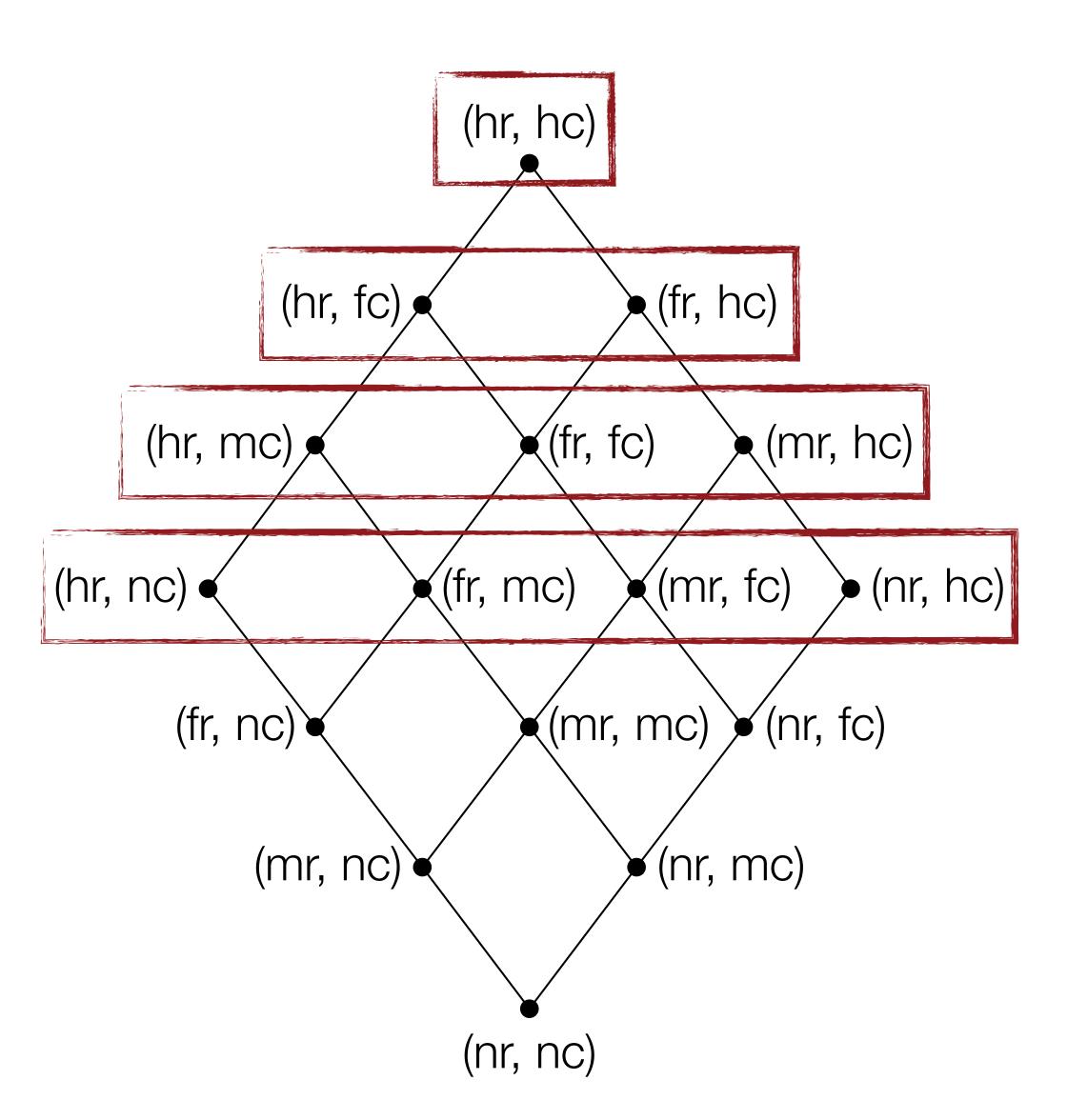


Skyline Order

• Skyline set: set of label vectors that are not worse than any other label vectors in all the dimensions;

$$l_a \preceq l'_a \quad \forall a \in A \quad \text{and} \quad \exists a_k \in A : l_{a_k} \prec l'_{a_k}$$

- $S_0 = \{(hr, hc)\}$
- $S_1 = \{(hr, fc), (fr, hc)\}$
- $S_2 = \{(hr, nc), (fr, mc), (mr, fc), (nr, hc)\}$
- ...



Theoretical Properties

- Utility oriented measures of retrieval effectiveness satisfy 2 properties:
- Replacement: if we replace a less relevant document with a more relevant one in the same rank position, the measure score should not decrease.
- Swap: if we swap a less relevant document in a higher rank position with a more relevant one in a lower rank position, the measure score should not decrease.







About 728.000.000 results (0.55 seconds)

https://www.bbc.com > news > health-56180921

vitamin D covid 19

Google

4 Apr 2021 — As Covid-19 swept the world, so did misinformation about how to treat it. But sometimes misinformation can develop even around ideas that ...

https://www.wcax.com > content > news > Research-sugge...

Research suggests vitamin D plays role in COVID-19 death ...

11 May 2020 — Vitamin D is key for maintaining healthy bones, and not having enough can affect the immune system and inflammation. Now, new research from ...

https://www.henryford.com > vitamin-d-and-covid-19 ▼

Will Boosting Your Vitamin D Intake Help Protect Against ...

5 Apr 2021 — There are many ways that vitamin D is good for you, but how about when it comes to COVID-19? Get the truth about the benefits of vitamin D.

https://globalnews.ca > news > covid-19-vitamin-d-link ▼

Can vitamin D lower the risk of COVID-19? Here's what we ...

27 Mar 2021 — There is no clear evidence that low vitamin D levels increase the risk of COVID-19 illness, but some experts recommend use of supplements ...

https://www.medicinenet.com > article

20 Vitamins and Supplements To Boost Immune Health for ...

27 May 2021 — Because COVID-19 comes with cold and flu-like symptoms, Vitamins B, C and D, as well as zinc may be helpful in boosting your immune system and ...

Vitamin D · Vitamin D Deficiency Quiz · Ascorbic acid

https://www.nature.com > ... > articles

The effect of high-dose parenteral vitamin D3 on COVID-19 ...

by M Güven · 2021 — In many studies, vitamin D has been found to be low in COVID-19 patients. In this study, we aimed to investigate the relationship between ...



Labels





Theoretical Properties - Multi-aspect Measures

 Partial order among labels can be used instead of "more relevant" and "less relevant";

$$GT(d,t) \sqsubset GT(d',t) \iff$$

$$l_a \preceq_a l'_a \quad \forall a \in A$$

$$\text{and } \exists \overline{a} \in A \colon l_{\overline{a}} \prec_{\overline{a}} l'_{\overline{a}}$$

 TOMA satisfies swap and replacements → extend findings for IR measures.

