

SIMILARITY SEARCH FOR MATHEMATICS

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FINE-TUNING QUERY EXPANSION AND UNIFICATION STRATEGIES

INTRODUCTION

Masaryk University (MU) has entered the area of MIR during the development of the Czech Digital Mathematics Library (DML-CZ). It quickly became clear that Digital Mathematical Libraries (DMLs) are specific especially in handling of formulae.

MU has partnered in the development of the European Digital Mathematics Library (EuDML) and supports math formulae search as one of the math specific features. We have also paid attention to the user interface aspect: formulae in the query are rendered at the same time as the user writes it.

EuDML with Math Indexer and Searcher (MlaS) is the first digital library collecting non-born-digital PDFs that supports math search in full texts.

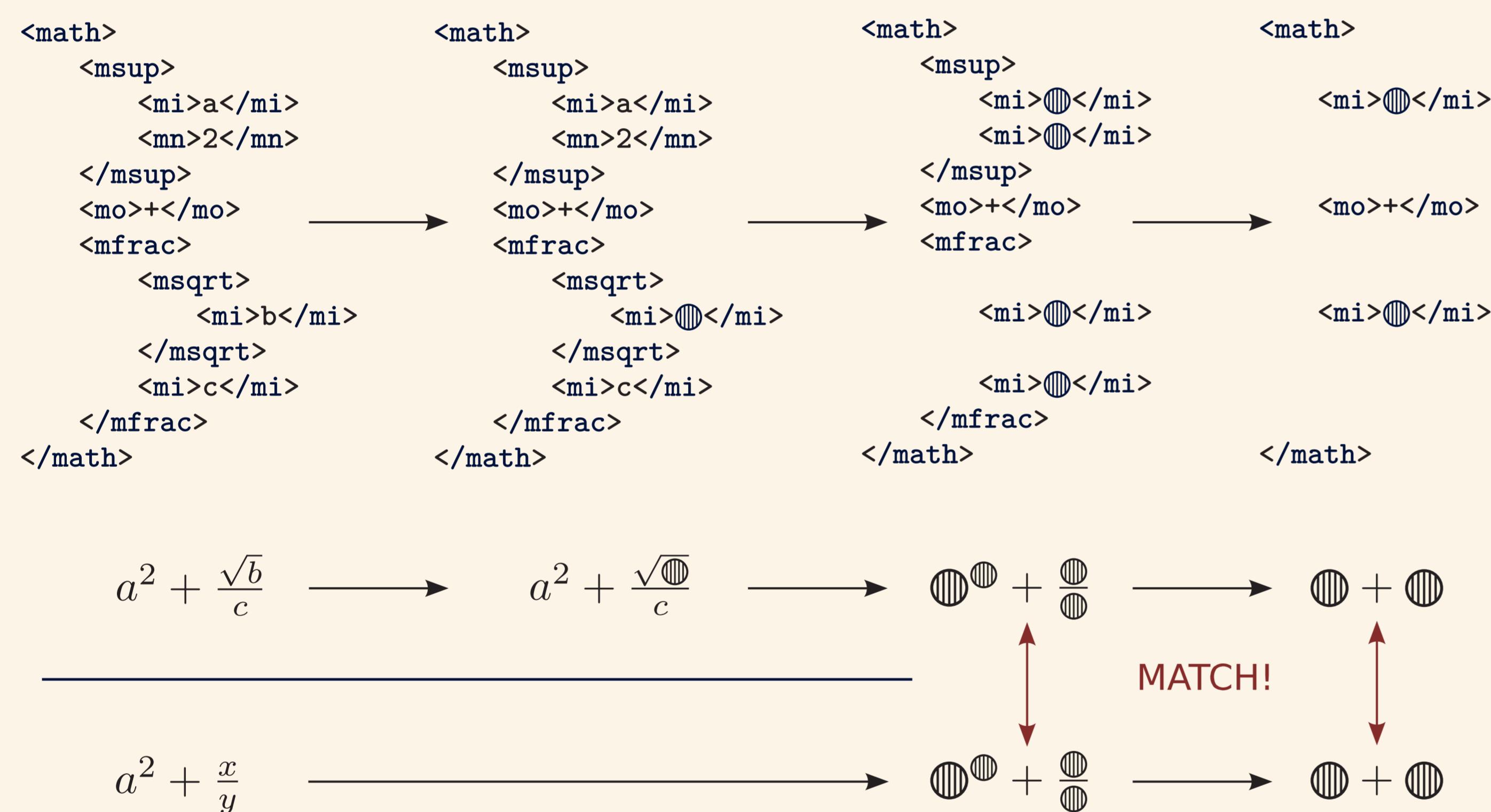
Our MIRMU team has been participating in NTCIR math information retrieval tasks since their introduction at NTCIR-10. This year we have tried three new approaches: structural unification, new querying strategies and new canonicalization procedures.

In our research we found out that structural unification increases recall but has negative impact on precision. NTCIR ground truth allowed us to compare effects of canonicalization and different querying strategies.

STRUCTURAL UNIFICATION

An important feature that we missed in NTCIR-11 was the ability to substitute structures. To structurally unify we implemented the open-source tool MathML Unifier (see <https://mir.fi.muni.cz/mathml-normalization/>) usable as a standalone command line utility or a Java library embeddable in other systems.

This tool is used for generating structurally unified versions of the input formulae. The unification is performed according to MathML tree layers.



FEATURES

Canonicalization The canonicalization process aims to normalize potential serializations (different notations in MathML encoding) of the same math formulae. The normalization is optimized for similarity search not to preserve full semantic information of the original formulae but possibly removes semantically negligible differences in behalf of similarity matches.

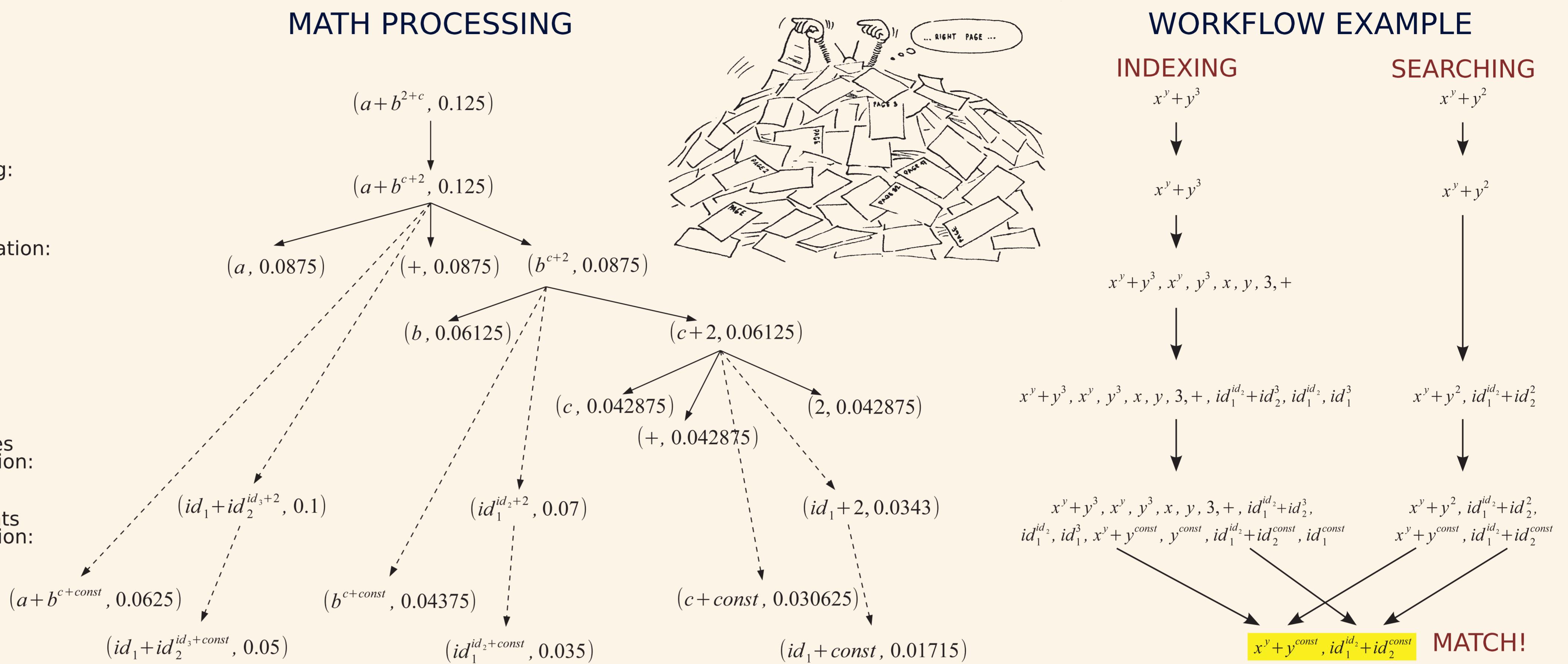
Canonicalization operators removal List of math operators to be removed from canonicalized formulae:

- U+2062 INVISIBLE TIMES
- U+22C5 DOT OPERATOR
- U+002A ASTERISK
- U+2063 INVISIBLE SEPARATOR
- U+2064 INVISIBLE PLUS

Unary operators removal Unary operators are removed from the input formulae in the process of formulae normalization by our MathML Canonicalizer.

Operator unification We define an operator equivalence relation. We substitute all of these operators with a canonical operator that represents each equivalence class.

Structural unification Indexing structurally unified derivatives of the original formulae was used for the first time by MlaS system at NTCIR-12.



The MathML Unifier tool is now integrated with our (Web)MlaS system and structural unification is done during indexing of formulae of the input documents as well as at query processing to structurally unify formulae from the user queries.

Proper weighting of structurally unified derivatives of the input formulae in relation to the original non-modified formulae and tokenized and unified subformulae is yet to be solved.

Unification is done from lists to the root of the MathML tree of the formula so that the substitution takes place for all the nodes in the given layer in one step and follows layer by layer up to the root. Final unification—to a single $\textcircled{+}$ —is omitted.

query 1 (the original query): $f_1 f_2 k_1 k_2 k_3$

query 2: $f_1 f_2 k_1 k_2$

query 3: $f_1 f_2 k_1$

query 4: $f_1 f_2$

query 5: $f_1 k_1 k_2 k_3$

query 6: $k_1 k_2 k_3$

QUERYING STRATEGIES

Based on NTCIR-11 ground truth of annotated data we developed an evaluation framework that allows us to rigorously compare several new querying strategies.

The query relaxation strategy used at NTCIR-11 we call **Leave Rightmost Out (LRO)**:

query 1 (the original query): $f_1 f_2 k_1 k_2 k_3$

query 2: $f_1 f_2 k_1 k_2$

query 3: $f_1 f_2 k_1$

query 4: $f_1 f_2$

query 5: $f_1 k_1 k_2 k_3$

query 6: $k_1 k_2 k_3$

Based on this concept we evaluated also other querying strategies:

Original Query Only (OQO) The basic reference querying strategy is to use the original query without any modifications or derived subqueries.

Math Terms Only (MTO) The query only consists of all formulae from the original query.

Text Terms Only (TTO) In Text Terms Only strategy the query only consists of text keywords from the original query.

All Possible Subqueries (APS) The opposite extreme to OQO is to use all potential subqueries derivable from the original query.

Leave One Out (LOO) This querying strategy is similar to the APS strategy with the following differences:

- We work with a restricted set of the subqueries—only the original query and derived subqueries with exactly one excluded component (one formula or one text keyword).

- Weight of interleaving ‘strips’ of results from subqueries is 2 if results are taken from the original query results list, and 1 otherwise.

Leave One or Two Out (LOOTO) The Leave One or Two Out querying strategy is a further extension of the previous Leave One Out strategy:

- The set of subqueries consists of the original query and derived subqueries with exactly one or two components excluded.
- The strip-weight is 3 if results are taken from the original query results list, 2 if results are taken from a derived query with exactly one excluded component, and 1 otherwise.

PHRASE EXPANSION

These modifications are only applicable on multi-word text keywords of the original query.

Original query

Formula 1: \textcircled{N}_0

Keyword 1: categorical simple theory

Phrase expansion For multi-word keywords individual words are used instead of the original multi-word keywords.

Formula 1: \textcircled{N}_0

Keyword 1: categorical

Keyword 2: simple

Keyword 3: theory

Full phrase expansion Individual words from the multi-word keywords are added one by one at the end of the keywords list (removing duplicates, if any).

Formula 1: \textcircled{N}_0

Keyword 1: categorical simple theory

Keyword 2: categorical

Keyword 3: simple

Keyword 4: theory

This modified version provides the querying strategy more flexibility for query relaxation and boolean operations on the query components.

CONCLUSIONS

The main advances of our approach since NTCIR-11 Math Task were development of our evaluation platform based on NTCIR-11 ground truth and introduction of math structural unification component as a part of the MlaS processing workflow.

Use of structurally unified derivatives increases recall but has negative impact on precision. Fine tuning the weights of structural unification nodes could possibly balance performance of our system towards recall at the expense of precision and vice versa. Setting and tuning the indexing and preprocessing parameters is necessary for given application.

We aim to reuse NTCIR-12 MathIR data as the ground truth in our evaluation platform to further improve performance of our system.

Our future MathIR research aims at incorporating machine learning techniques to formulae disambiguation and ranking, and deploying Computer Algebra Systems for better canonicalization of the input formulae.

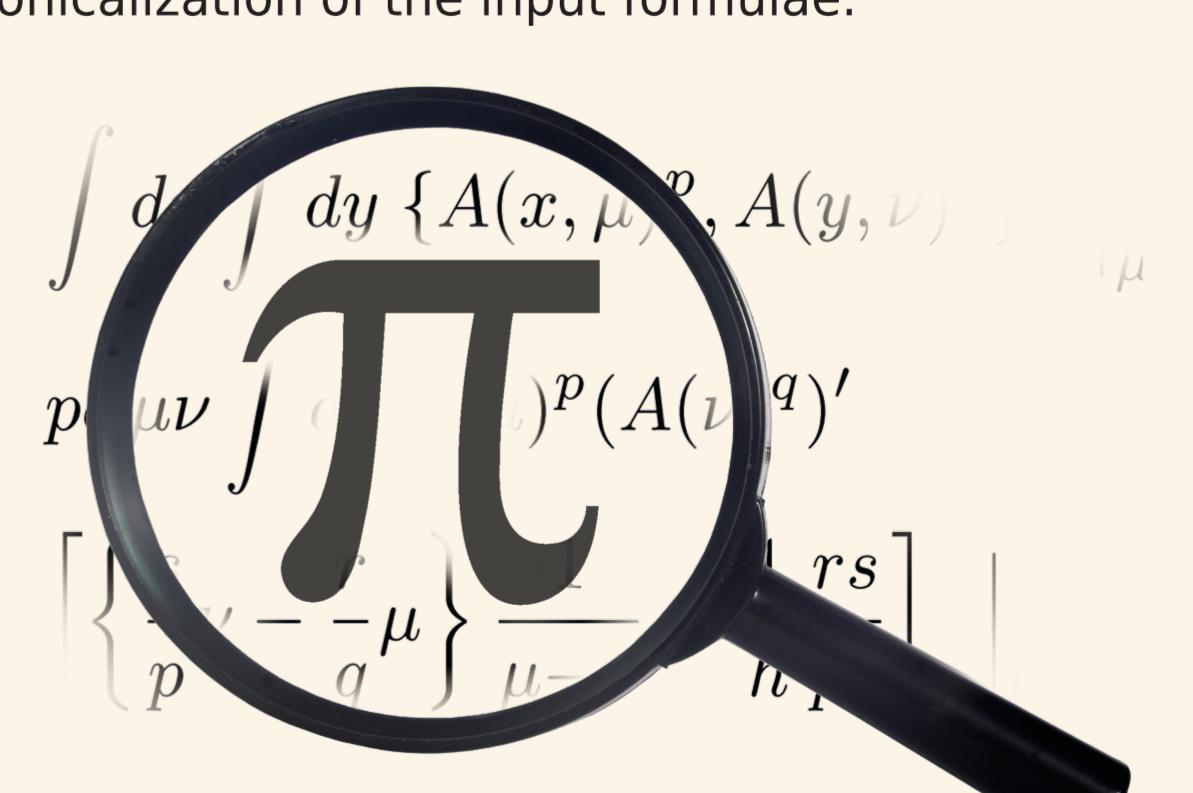


table). This gave us 660 results describing the performance of each particular combination. We used MAP and Bpref metrics for evaluation against NTCIR-11 ground truth.

For NTCIR-12 submission we have chosen four most promising or curious configurations in terms of Bpref and MAP.

Cell colours indicate groups of comparable combinations with respect to Bpref metric.

We have built several dozen indices for the Main and Wiki Math Task with different features and configurations enabled. Every row in the table corresponds to the particular combination of features.

We queried each index with full 50 topics from NTCIR-11 with 11 different querying strategies (columns in the

A	B	C	D	E	F	G	H	I	J	K	L	M
1	BPREF	PCM-T	CM-APSWMWWS	CM-LOOT	CM-LOO	CM-N1MIR	CM-N1MRMSF	CM-N1MRMSF	CM-N1MIR	CM-QFPQ	CM-QKPO	CM-OQO
4	ntcir-12-10-pfe	0.0492	0.1029	0.1011	0.0987	0.1047	0.1047	0.1057	0.1027	0.1055	0.0220	0.0619
5	ntcir-12-11	0.0557	0.0897	0.0873	0.0898	0.0972	0.0977	0.0973	0.0956	0.0977	0.0216	0.0514
6	ntcir-12-11-fe	0.0529	0.0964	0.0991	0.1040	0.1105	0.1104	0.1107	0.1079	0.1110	0.0216	0.0673
7	ntcir-12-11-pfe	0.0509	0.1037	0.1015	0.0987	0.1045	0.1038	0.1055	0.1019	0.1051	0.0216	0.0616
8	ntcir-12-12	0.0552	0.0893	0.0866	0.0891	0.0970	0.0971	0.0969	0.0951	0.0969	0.0226	0.0511
9	ntcir-12-12-fe	0.0527	0.0949	0.0969	0.1023	0.1094	0.1090	0.1097	0.1073	0.1094	0.0226	0.0670
10	ntcir-12-12-pfe	0.0502	0.1009	0.0989	0.0976	0.1040	0.1033	0.1045	0.1015	0.1040	0.0226	0.0614
11	ntcir-12-13	0.0561	0.0882	0.0860	0.0889	0.0960	0.0967	0.0962	0.0945	0.0970	0.0224	0.0513
12	ntcir-12-13-fe	0.0529	0.0963	0.0981	0.1032	0.1089	0.1096	0.1094	0.1067	0.1099	0.0224	0.0671
13	ntcir-12-13-pfe	0.0517	0.1014	0.1012	0.0986	0.1035	0.1034	0.1043	0.1013	0.1045	0.0224	0.0616
14	ntcir-12-14	0.0548	0.0871	0.0858	0.0872	0.0944	0.0945	0.0947	0.0927	0.0947	0.0221	0.0511
15	ntcir-12-14-fe	0.0517	0.0935	0.0954	0.1008	0.1066	0.1071	0.1078	0.1046	0.1075	0.0221	0.0668
16	ntcir-12-14-pfe	0.0503	0.0995	0.0988	0.0965	0.1021	0.1014	0.1029	0.1096	0.1024	0.0221	0.0613
17	ntcir-12-15	0.0965	0.0892	0.0870	0.0895	0.0984	0.0986	0.0984	0.0966	0.0987	0.0220	0.0515
18	ntcir-12-15-fe	0.1067	0.0942	0.0980	0.1033	0.1096	0.1102	0.1102	0.1081	0.1108	0.0220	0.0671
19	ntcir-12-15-pfe	0.1021	0.1013	0.1001	0.0979	0.1044	0.1040	0.1052	0.1020	0.1048	0.0220	0.0614
20	ntcir-12-16	0.0955	0.0884	0.0870	0.0895	0.0974	0.0976	0.0980	0.0961	0.0974	0.0225	0.0512
21	ntcir-12-16-fe	0.1058	0.0961	0.0988	0.1038	0.1097	0.1100	0.1104	0.1079	0		