

Structural representation of result: not only reachability, but paths; debuggibg; understanding.

3 PARSER-COMBITATORS FOR PATH QUERYING

In this section we present our implementation of and describe some details.

Our implemenetation is based on Meerkat library. We need only some steps for generalization.

As far as linear input is a one of case of graph, it is possible to provide input abstraction which make possible to generalize combinators.

SPPF may be an arbitrary graph in opposite of linear input parsing.

Let we introduce an example. Graph. Grammar. In terms of combinators.

- Interface for Neo4j bata base
- Extensible solution
- An architecture of the solution.

4 EVALUATION

In this section we present evaluation of Meerkat grph querying library. We show its perfomance on a classical ontology graphs for in memory graph and for Neo4j database, show application on may-alias static code analysis problem, and compare with Trails [9] library for graph traversals.

All tests are perfomed on a machine running Fedora 27 with quad-core Intel Core i7 2.5 GHz CPU with 8 GB of memory.

4.1 Ontology querying

One of well-known graph querying problems is a queries for ontologies [1]. We use Meerkat to evaluate it on some popular ontologies presented as RDF files from paper [19]. We convert RDF files to a labeled directed graph like the following: for every RDF triple (*subject*, *predicate*, *object*) we create two edges (*subject*, *predicate*, *object*) and (*object*, *predicate*⁻¹, *subject*). On those graphs we apply two queries from the paper [5] which grammars are in Fig. 1, and Fig. 2

$$\begin{aligned}
 S &\rightarrow subClassOf^{-1} S subClassOf \\
 S &\rightarrow type^{-1} S type \\
 S &\rightarrow subClassOf^{-1} subClassOf \\
 S &\rightarrow type^{-1} type
 \end{aligned}$$

Figure 1: Query 1 grammar

Meerkat represetaion of those queries in terms of parser combinators is similar to its EBNF form and presnted in Fig. 3 and Fig. 4

The queries applied in two following ways.

- Convert RDF files to a graph input for meerkat and then directly parse on query 1 and query 2
- Convert RDF files to a Neo4j database and then parse this database on given queries

$$\begin{aligned}
 S &\rightarrow B subClassOf \\
 B &\rightarrow subClassOf^{-1} B subClassOf \\
 B &\rightarrow subClassOf^{-1} subClassOf
 \end{aligned}$$

Figure 2: Query 2 grammar

```

val S: Nonterminal = syn(
  "subclassof-1" ~~ S ~~ "subclassof" |
  "type-1" ~~ S ~~ "type" |
  "subclassof-1" ~~ "subclassof" |
  "type-1" ~~ "type")

```

Figure 3: Meerkat representation of Query 1

```

val S: Nonterminal = syn(
  "subclassof-1" ~~ S ~~ "subclassof" |
  "subclassof")

```

Figure 4: Meerkat representation of Query 2

Table 1 shows experimental results of those two aproaches over the testing RDF files where *number of results* is a number of pairs of nodes (v_1 , v_2) such that exists S-path from v_1 to v_2 .

The performance is about 2 times slower than in [5] and shows the same results. If compare the performance of in memory graph querying and database querying, the second one is slower in about 2 – 4 times.

4.2 Integration with Neo4j

4.3 Static code analysis

Alias analysis is one of the fundamental static analysis problems [10]. Alias analysis checks may-alias relations between code expressions and can be formulated as a Context-Free language (CFL) reachability problem [14]. In that case program represeted as Program Expression Graph (PEG) [20]. Verticies in PEG are program expressions and edges are relations between them. In a case of analysing C source code there is two kind of edges **D**-edge and **A**-edge.

- Pointer dereference edge (**D**-edge). For each pointer deference $*e$ there is a directed D-edge from e to $*e$.
- Pointer assignment edge (**A**-edge). For each assignment $*e_1 = e_2$ there is a directed A-edge from e_2 to $*e_1$

Also, for the sake of simplicity, there are edges labeled by \bar{D} and \bar{A} which corresponds to reversed D-edge and A-edge, respectively.

The grammar for may-alias problem from [20] presented in Fig. 5. It consists of two nonterminals **M** and **V**. It allows us to make two kind of queries for each of nonterminals **M** and **V**.

- **M** production shows that two l-value expression are memory aliases i.e. may stands for the same memory location.
- **V** shows that two expression are value aliases i.e. may evaluate to the same pointer value.

Ontology	#tripples	Query 1				Query 2			
		#results	In memory graph (ms)	DB query time (ms)	Trails (ms)	#results	In memory graph (ms)	DB query time (ms)	Trails (ms)
atom-primitive	425	15454	174	236	2849	122	49	56	453
biomedical-mesure-primitive	459	15156	328	398	3715	2871	36	52	60
foaf	631	4118	23	42	432	10	1	2	1
funding	1086	17634	151	175	367	1158	18	23	76
generations	273	2164	9	27	9	0	0	0	0
people_pets	640	9472	68	87	75	37	2	3	2
pizza	1980	56195	711	792	7764	1262	44	56	905
skos	252	810	4	29	6	1	0	1	0
travel	277	2499	23	93	34	63	2	2	1
univ-bench	293	2540	19	74	31	81	2	3	2
wine	1839	66572	578	736	3156	133	5	7	4

Table 1: Evaluation results for In Memory Graph and Graph DB

$$M \rightarrow \bar{D} V D$$

$$V \rightarrow (M? A)^* M? (A M?)^*$$

Figure 5: Context-Free grammar for the may-alias problem in syntax

```

val B = (out("type-1") ~> out("type")) |
  (out("subclassof-1") ~> B ~> out("subclassof")) |
  (out("type-1") ~> B ~> out("type")) |
  (out("subclassof-1") ~> out("subclassof"))
val S = V ~ B

```

Figure 6: Trails representation of Query 1

Program	Code Size (KLOC)	Count of aliases		Time (ms)
		M aliases	V aliases	
wc-5.0	0.5K	0	174	350
pr-5.0	1.7K	13	1131	532
ls-5.0	2.8K	52	5682	436
bzip2-1.0.6	5.8K	9	813	834
gzip-1.8	31K	120	4567	1585

Table 2: Running may-alias queries on Meerkat on some C open-source projects

```

val B = out("subclassof") |
  (out("subclassof-1") ~> B ~> out("subclassof"))
val S = V ~ B

```

Figure 7: Trails representation of Query 2

We made **M** and **V** queries on the code some open-source C projects. The results are presented on the Table 2

It may be usefull for tools development.

4.4 Comparison with GLL

4.5 Comparison with Trails

Trails [9] is a Scala graph combinator library which provides combinators for graph traversals. It provides traversers for describing paths in graphs in terms of parser combinators and allows to get results as a stream (maybe infinite) of all possible paths described by composition of basic traversals. Trails as well as Meerkat support parsing in memory graphs, so we compare performance of Trails and Meerkat on the queries from subsection 4.1. Query 1 and Query 2 in terms of Trails are in Fig. 6 and Fig. 7.

Here queries made in same way as in Meerkat. S traversal returns pairs (*begin node*, *start node*) of S-path. Combine operators ~ and ~> in queries made in the way to get only last node from path.

The result of comparison are in table 1. Trails gives the same results as Meerkat (column *results* in table 1) but slower than Meerkat.

1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1

5 CONCLUSION

We propose a native way to integrate language for language-constrained path querying into general purpose programming language. We implement it and show that our implementation can be applied for real problems.

Code is available on GitHub:

Future work is

SPPF processing for debugging and results processing

Attributed grammars processing to provide mechanim for semantics calualtion

REFERENCES

- [1] Serge Abiteboul, Richard Hull, and Victor Vianu. 1995. Foundations of Databases. (1995).
- [2] Pablo Barceló, Gaelle Fontaine, and Anthony Widjaja Lin. 2013. Expressive Path Queries on Graphs with Data. In *International Conference on Logic for Programming Artificial Intelligence and Reasoning*. Springer, 71–85.
- [3] Chris Barrett, Riko Jacob, and Madhav Marathe. 2000. Formal-language-constrained path problems. *SIAM J. Comput.* 30, 3 (2000), 809–837.

- [4] James Cheney, Sam Lindley, and Philip Wadler. 2013. A Practical Theory of Language-integrated Query. *SIGPLAN Not.* 48, 9 (Sept. 2013), 403–416. <https://doi.org/10.1145/2544174.2500586>
- [5] Semyon Grigorev and Anastasiya Ragozina. 2016. Context-Free Path Querying with Structural Representation of Result. *CoRR* abs/1612.08872 (2016). <http://arxiv.org/abs/1612.08872>
- [6] Jelle Hellings. 2014. Conjunctive context-free path queries. (2014).
- [7] Jelle Hellings. 2015. Path Results for Context-free Grammar Queries on Graphs. *CoRR* abs/1502.02242 (2015). <http://arxiv.org/abs/1502.02242>
- [8] Anastasia Izmaylova, Ali Afroozeh, and Tijs van der Storm. 2016. Practical, General Parser Combinators. In *Proceedings of the 2016 ACM SIGPLAN Workshop on Partial Evaluation and Program Manipulation (PEPM '16)*. ACM, New York, NY, USA, 1–12. <https://doi.org/10.1145/2847538.2847539>
- [9] Daniel Kröni and Raphael Schweizer. 2013. Parsing Graphs: Applying Parser Combinators to Graph Traversals. In *Proceedings of the 4th Workshop on Scala (SCALA '13)*. ACM, New York, NY, USA, Article 7, 4 pages. <https://doi.org/10.1145/2489837.2489844>
- [10] Thomas J. Marlowe, William G. Landi, Barbara G. Ryder, Jong-Deok Choi, Michael G. Burke, and Paul Carini. 1993. Pointer-induced Aliasing: A Clarification. *SIGPLAN Not.* 28, 9 (Sept. 1993), 67–70. <https://doi.org/10.1145/165364.165387>
- [11] Erik Meijer, Brian Beckman, and Gavin Bierman. 2006. LINQ: Reconciling Object, Relations and XML in the .NET Framework. In *Proceedings of the 2006 ACM SIGMOD International Conference on Management of Data (SIGMOD '06)*. ACM, New York, NY, USA, 706–706. <https://doi.org/10.1145/1142473.1142552>
- [12] A. Mendelzon and P. Wood. 1995. Finding Regular Simple Paths in Graph Databases. *SIAM J. Computing* 24, 6 (1995), 1235–1258.
- [13] Joan Gerard Rekers. 1992. *Parser generation for interactive environments*. Ph.D. Dissertation. Universiteit van Amsterdam.
- [14] Thomas Reps. 1997. Program Analysis via Graph Reachability. In *Proceedings of the 1997 International Symposium on Logic Programming (ILPS '97)*. MIT Press, Cambridge, MA, USA, 5–19. <http://dl.acm.org/citation.cfm?id=271338.271343>
- [15] Juan L Reutter, Miguel Romero, and Moshe Y Vardi. 2015. Regular queries on graph databases. *Theory of Computing Systems* (2015), 1–53.
- [16] Elizabeth Scott and Adrian Johnstone. 2010. GLL parsing. *Electronic Notes in Theoretical Computer Science* 253, 7 (2010), 177–189.
- [17] Petteri Sevon and Lauri Eronen. 2008. Subgraph queries by context-free grammars. *Journal of Integrative Bioinformatics* 5, 2 (2008), 100.
- [18] Mihalis Yannakakis. 1990. Graph-theoretic methods in database theory. In *Proceedings of the ninth ACM SIGACT-SIGMOD-SIGART symposium on Principles of database systems*. ACM, 230–242.
- [19] Xiaowang Zhang, Zhiyong Feng, Xin Wang, Guozheng Rao, and Wenrui Wu. 2016. Context-free path queries on RDF graphs. In *International Semantic Web Conference*. Springer, 632–648.
- [20] Xin Zheng and Radu Rugina. 2008. Demand-driven Alias Analysis for C. In *Proceedings of the 35th Annual ACM SIGPLAN-SIGACT Symposium on Principles of Programming Languages (POPL '08)*. ACM, New York, NY, USA, 197–208. <https://doi.org/10.1145/1328438.1328464>