

Incremental Maintenance of Graph Views

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March 9, 2021

This Talk

Our Work

A machine-verified *incremental* graph query engine.

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A [machine-verified](#) *incremental* graph query engine.

- ...relying on the [Coq proof assistant](#)

This Talk

Our Work

A machine-verified *incremental* graph query engine.

- ...relying on the Coq proof assistant
- ...for *regular queries*¹ (closure operators as primitive)
 - expressive & tractable²
 - ⇒ amenable to *graph querying*
 - logic-based language
 - ⇒ amenable to *formal verification*

¹ Moshe Y. Vardi: *A Theory of Regular Queries*. PODS 2016: 1-9

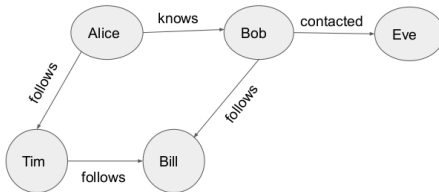
² Juan L. Reutter, et al.: *Regular Queries on Graph Databases*. Th. Comput. Syst. 61(1): 31-83 (2017)

Datalog

- logic programming without function symbols
- consists of **facts** and **rules**
- **finite model semantics**
- **sound**, **complete** and **terminating** evaluation
→ *bottom-up iteration of a fixpoint operator*

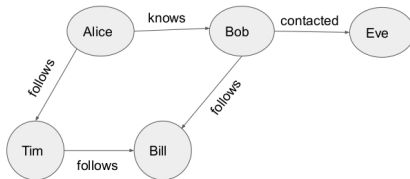
Datalog Example

Assume we have the following simple graph database instance:



...and that we want to compute *all the pairs of potential friends*.

Datalog Example



```

knows(Alice,Bob)
contacted(Bob,Eve)
follows(Alice,Tim)
follows(Tim,Bill)
follows(Bob,Bill)

```

```

pfriends(X,Y) ← knows(X,Y), connected(X,Z), connected(Y,Z)

```

```

pfriends(X,Y) ← contacted(X,Y)

```

```

connected(X,Y) ← follows(X,Z), connected(Z,Y)

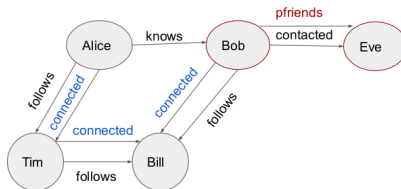
```

```

connected(X,Y) ← follows(X,Y)

```

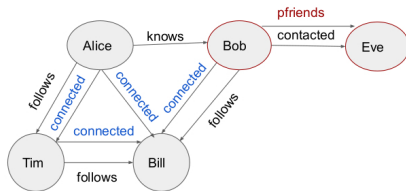
Datalog Example



knows(Alice,Bob) ☒
 contacted(Bob,Eve) ☒ **pfriends**(Bob,Eve)
 follows(Alice,Tim) ☒ **connected**(Alice,Tim)
 follows(Tim,Bill) ☒ **connected**(Tim,Bill)
 follows(Bob,Bill) ☒ **connected**(Bob,Bill)

pfriends(X,Y) ← knows(X,Y), **connected**(X,Z), **connected**(Y,Z)
pfriends(X,Y) ← contacted(X,Y) ☒
connected(X,Y) ← follows(X,Z), **connected**(Z,Y)
connected(X,Y) ← follows(X,Y) ☒

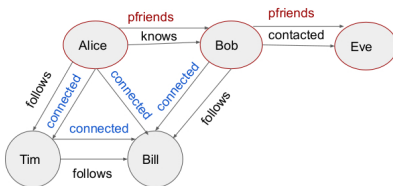
Datalog Example



knows(Alice,Bob)
 contacted(Bob,Eve)
 follows(Alice,Tim) ☒
 follows(Tim,Bill)
 follows(Bob,Bill)
 pfriends(Bob,Eve)
 connected(Alice,Tim)
 connected(Tim,Bill) ☒
 connected(Alice,Bill)
 connected(Bob,Bill)

pfriends(X,Y) ← knows(X,Y), connected(X,Z), connected(Y,Z)
 pfriends(X,Y) ← contacted(X,Y)
 connected(X,Y) ← follows(X,Z), connected(Z,Y) ☒
 connected(X,Y) ← follows(X,Y) ☒

Datalog Example



knows(Alice,Bob) ☒

contacted(Bob,Eve)

follows(Alice,Tim)

follows(Tim,Bill)

follows(Bob,Bill)

pfriends(Bob,Eve)

connected(Alice,Tim)

connected(Tim,Bill)

connected(Bob,Bill) ☒

connected(Alice,Bill) ☒

pfriends(Alice,Bob)

pfriends(X,Y) ← knows(X,Y), connected(X,Z), connected(Y,Z) ☒

pfriends(X,Y) ← contacted(X,Y)

connected(X,Y) ← follows(X,Z), connected(Z,Y)

connected(X,Y) ← follows(X,Y)

Regular Datalog

- **safe**: head variables in each clause appear in its body
- recursion is restricted to **transitive closure**
...and internalized into special literals
- distinguished top clause whose head we call **view**
- all symbols (except the view one) are **binary**
- **stratified**: no clauses with body symbols not previously defined

Datalog vs. Regular Datalog

```

pconnected(X,Y) ← pfriends(X,Y)
pconnected(X,Y) ← pfriends(X,Z), pconnected(Z,Y)
pfriends(X,Y) ← knows(X,Y), connected(X,Z), connected(Y,Z)
pfriends(X,Y) ← contacted (X,Y)
connected(X,Y) ← follows(X,Z), connected(Z,Y)
connected(X,Y) ← follows(X,Y)

```



```

pconnected(X,Y) ← pfriends*(X,Y)
pfriends(X,Y) ← ((knows ∧ (connected · connected-)) ∨ contacted) (X,Y)
connected(X,Y) ← follows*(X,Y)

```

- Regular Datalog corresponds to **UC2RPQ under transitive closure**
 - expressive: allows for complex graph patterns
 - tractable: parallelizability & decidable query containment

Building Blocks

Graph Databases

finite sets of constants $\mathbf{V}$ (nodes) & symbols Σ (edge labels).

Graph Instance $\mathcal{G}$ over Σ :

set of *directed* labeled edges, $\mathbf{E}$, where $\mathbf{E} \subseteq \mathbf{V} \times \Sigma \times \mathbf{V}$.

Graph Database $\mathcal{D}(\mathcal{G})$ over $\mathcal{G}$:

$\mathcal{G}$ can be seen as a database $\mathcal{D}(\mathcal{G}) = \{s(n_1, n_2) \mid (n_1, s, n_2) \in \mathbf{E}\}$

Path ρ of length k in $\mathcal{G}$: sequence $n_1 \xrightarrow{s_1} n_2 \dots n_{k-1} \xrightarrow{s_k} n_k$

Path Label: $\lambda(\rho) = s_1 \dots s_k \in \Sigma^*$

Syntax

Regular Datalog (RD) Expressions

Terms (Node IDs)	$t ::= x \mid n$	where $x \in Var, n \in \mathbf{V}$
Atoms	$A ::= s(t_1, t_2)$	where $s \in \Sigma$
Literals	$L ::= A \mid A^+$	
Conjunctive Body	$B ::= L_1 \wedge \dots \wedge L_n$	where $n \in \mathbb{N}$
Disjunctive Body	$D ::= B_1 \vee \dots \vee B_n$	where $n \in \mathbb{N}$
Clauses	$C ::= (t_1, t_2) \leftarrow D$	
Programs	$\Pi : \Sigma \rightarrow \{C_1, \dots, C_n\}$	where $n \in \mathbb{N}$

Regular Queries (RQ) over $\mathcal{G}$

- RD-program Π
- distinguished **query clause** Ω whose head is the top-level **view**

Semantics (I/II)

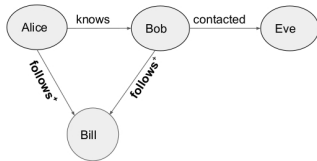
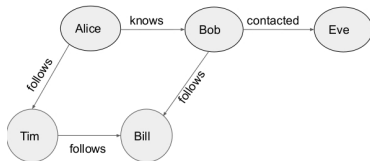
Interpretations ($\mathcal{G}$)

Modeled as *indexed relations* $(\Sigma \times \{\square, +\}) \rightarrow \mathcal{P}(\mathbf{V} \times \mathbf{V})$.

Semantics (I/II)

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Semantics (I/II)

Interpretations ($\mathcal{G}$)

Modeled as *indexed relations* $(\Sigma \times \{\square, +\}) \rightarrow \mathcal{P}(\mathbf{V} \times \mathbf{V})$.

Interpretation Well-Formedness (wfG)

$\mathcal{G}(s, +)$ has to correspond to the transitive closure of $\mathcal{G}(s, \square)$:

$$\text{wfG}(\mathcal{G}) \iff \forall s, \text{is_closure}(\mathcal{G}(s, \square), \mathcal{G}(s, +))$$

$$\text{is_closure}(g_s, g_p) \iff \forall (n_1, n_2) \in g_p, \exists \rho \in \mathbf{V}^+, \text{path}(g_s, n_1, \rho) \wedge \text{last}(\rho) = n_2$$

$$\text{path}(g, n_1, \rho) \iff \forall i \in \{1 \dots |\rho|\}, (n_i, n_{i+1}) \in g$$

Semantics (II/II)

Literal Satisfaction

For $L \equiv s^l(n_1, n_2)$, $\mathcal{G} \models L \iff (n_1, n_2) \in \mathcal{G}(s, l)$.

Clause Satisfaction

For $C \equiv (t_1, t_2) \leftarrow (L_{1,1} \wedge \dots \wedge L_{1,n}) \vee \dots \vee (L_{m,1} \wedge \dots \wedge L_{m,n})$,
 $\mathcal{G} \models_s L \iff \forall \eta, \forall_{i=1..m} (\bigwedge_{j=1..n} \mathcal{G} \models \eta(L_{i,j})) \Rightarrow \mathcal{G} \models \eta(s(t_1, t_2))$.

Program Satisfaction

For $\Pi \equiv \Sigma \rightarrow \{C_1, \dots, C_n\}$, $\mathcal{G} \models_\Sigma \Pi \iff \forall s \in \Sigma, \mathcal{G} \models_s \Pi(s)$.

View Computation: Clausal Evaluation

For each clause $C \equiv (t_1, t_2) \leftarrow \bigvee_{i=1..n} B_i$ corresponding to $\Pi(s)$

- match each conjunctive body B_i against the interpretation
 - extend substitutions $M_G^A(a)$ matching a given atom a
 - ...uniformly, to the entire body as $M_G^B(B_i)$
- collect the substitutions matching the entire disjunctive body

Clausal Consequence Operator

$$\mathcal{T}^{\Pi, s}(\mathcal{G}) \equiv \{\sigma(t_1, t_2) \mid \sigma \in \bigcup_{i=1..n} M_G^B(B_i)\}$$

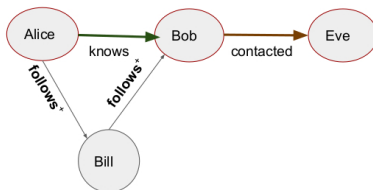
View Computation: Stratified Program Evaluation

Stratification Condition

A program Π is *stratified* if: there exists a mapping $\sigma : \Sigma \rightarrow [1, n]$ such that, for all s in Σ , the $\Pi(s)$ clause $(t_1, t_2) \leftarrow B$ satisfies:

$$\max_{r \in \text{sym}(B)} \sigma(r) < \sigma(s)$$

View Computation: Stratified Program Evaluation

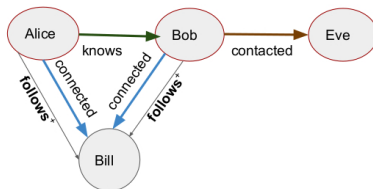


$$\Pi_1 = \{ \text{connected}(X, Y) \leftarrow \text{follows}^+(X, Y) \}$$

$$\Pi_2 = \{ \text{pfriends}(X, Y) \leftarrow ((\text{knows} \wedge (\text{connected} \cdot \text{connected}^-)) \vee \text{contacted})(X, Y) \}$$

$$\Pi_3 = \{ \text{pconnected}(X, Y) \leftarrow \text{pfriends}^+(X, Y) \}$$

View Computation: Stratified Program Evaluation



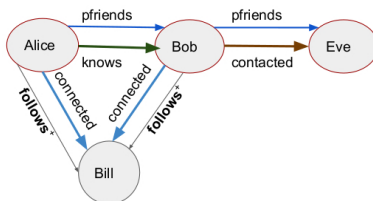
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$$\Delta_1 = T^{\Pi_1, \text{connected}}(\emptyset) = \text{connected} \rightarrow \{(Alice, Bill), (Bob, Bill)\}$$

View Computation: Stratified Program Evaluation



$$\Pi_1 = \{ \text{connected}(X, Y) \leftarrow \text{follows}^+(X, Y) \}$$

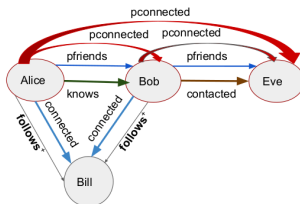
$$\Pi_2 = \{ \text{pfriends}(X, Y) \leftarrow ((\text{knows} \wedge (\text{connected} \cdot \text{connected}^-)) \vee \text{contacted})(X, Y) \}$$

$$\Pi_3 = \{ \text{pconnected}(X, Y) \leftarrow \text{pfriends}^+(X, Y) \}$$

$$\Delta_1 = T^{\Pi_1, \text{connected}}(\emptyset) = \text{connected} \rightarrow \{(Alice, Bill), (Bob, Bill)\}$$

$$\Delta_2 = T^{\Pi_2, \text{pfriends}}(\Delta_1) = \text{pfriends} \rightarrow \{(Alice, Bob), (Bob, Eve)\}$$

View Computation: Stratified Program Evaluation



$$\Pi_1 = \{ \text{connected}(X, Y) \leftarrow \text{follows}^+(X, Y) \}$$

$$\Pi_2 = \{ \text{pfriends}(X, Y) \leftarrow ((\text{knows} \wedge (\text{connected} \cdot \text{connected}^-)) \vee \text{contacted})(X, Y) \}$$

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$$\Delta_1 = T^{\Pi_1, \text{connected}}(\emptyset) = \text{connected} \rightarrow \{(Alice, Bill), (Bob, Bill)\}$$

$$\Delta_2 = T^{\Pi_2, \text{pfriends}}(\Delta_1) = \text{pfriends} \rightarrow \{(Alice, Bob), (Bob, Eve)\}$$

$$\Delta_3 = T^{\Pi_3, \text{pconnected}}(\Delta_1 \cup \Delta_2) = \text{pconnected} \rightarrow \{(Alice, Bob), (Bob, Eve), (Alice, Eve)\}$$

Modeling Updates

Updates

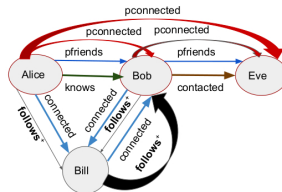
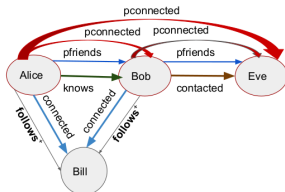
An *update* $\Delta \equiv (\Delta_+, \Delta_-)$ is a pair of *disjoint* graphs Δ_+, Δ_- .

- $\Delta_+ \equiv$ bulk insertions
- $\Delta_- \equiv$ bulk deletions

Update Application

$$\begin{aligned}\mathcal{G} :+ \Delta &\equiv \mathcal{G} \setminus \Delta_- \cup \Delta_+ \\ \Delta\{s \rightarrow (g_+, g_-)\} &\equiv (\Delta_+\{s \rightarrow g_+\}, \Delta_-\{s \rightarrow g_- \setminus g_+\})\end{aligned}$$

View Maintenance Computation



$$\Pi_1 = \{ \text{connected}(X, Y) \leftarrow \text{follows}^+(X, Y) \}$$

$$\Pi_2 = \{ \text{pfriends}(X, Y) \leftarrow ((\text{knows} \wedge (\text{connected} \cdot \text{connected}^-)) \vee \text{contacted})(X, Y) \}$$

$$\Pi_3 = \{ \text{pconnected}(X, Y) \leftarrow \text{pfriends}^+(X, Y) \}$$

$$\Delta'_1 = T_{\text{supp}}^{\Pi_1, \text{connected}}(\Delta_1) = \Delta_1 \{ \text{connected} \rightarrow (\{(Bill, Bob)\}, \emptyset) \}$$

... where $\text{supp} = \{ \text{knows}, \text{contacted}, \text{follows}^+, \text{connected}, \text{pconnected} \}$

View Computation vs. Maintenance

For each clause $C \equiv (t_1, t_2) \leftarrow \bigvee_{i=1..n} B_i$ corresponding to $\Pi(s)$

Incremental Clausal Consequence Operator

$$T_{\mathcal{G}, \text{supp}}^{\Pi, s}(\Delta) = \begin{cases} T^{\Pi, s}(\mathcal{G} \text{ :+ : } \Delta), & (s \notin \text{supp}) \vee (\Delta_- \cup D) \neq \emptyset \\ \bigcup_{i=1..n} M_{\mathcal{G}, \Delta}^B(B_i), & \text{otherwise} \end{cases}$$

Full Engine Overview

- stratified, single-pass, **bottom-up heuristic**
- **non-recursive** (recursion internalized in closure computation)
- supports both view **computation** and **incremental maintenance**
- core component: **clause evaluation**
 - *forward-chain clausal consequence operator* (`fwd_or_clause`)
 - based on a *matching algorithm*
 - corresponds to computing a *nested-loop join*

Full Engine Overview

```
Fixpoint fwd_program  $\Pi$   $\mathcal{G}$  supp  $\Delta$   $\Sigma_{\triangleright}$   $\Sigma_{\triangleleft}$  : edelta :=  
  match  $\Sigma_{\triangleleft}$  with  
  | [::]           =>  $\Delta$   
  | [:: s & ss] =>  
    let (arg, body) :=  $\Pi$  s                                in  
    let  $\Delta'$  := fwd_or_clause  $\mathcal{G}$  supp  $\Delta$  s arg body in  
    let  $\Delta'$  := compute_closures  $\mathcal{G}$   $\Delta'$  s              in  
    fwd_program  $\Pi$   $\mathcal{G}$  supp  $\Delta'$  ( $s \cup s^+ \cup \Sigma_{\triangleright}$ ) ss
```

Regular Datalog: Engine Characterization

Theorem (Soundness)

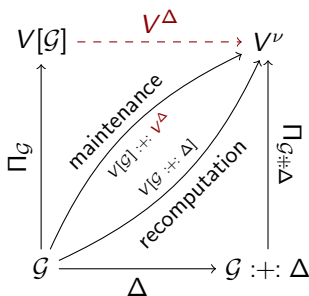
- Π – a safe, stratifiable, Regular Datalog program
- Σ – its set of symbols
- $\mathcal{G}$ – a graph instance
- Δ – an update

The IVM-engine cumulatively processes symbols in Σ , such that if:

- the already processed symbols, $\Sigma_{\triangleright}$, are a well-formed Π -slice
- Δ only modifies $\Sigma_{\triangleright}$, i.e., $\text{sym}(\Delta) \subseteq \Sigma_{\triangleright}$
- $\mathcal{G} :+ \Delta \models_{\Sigma_{\triangleright}} \Pi$

Then, it outputs Δ_o , such that $\mathcal{G} :+ \Delta_o \models_{\Sigma} \Pi$.

Incremental View Maintenance (IVM)



$\mathcal{G} \equiv$ base graph; $\Pi \equiv$ RD program; $V \equiv$ top-view; $\Delta \equiv$ update.

Soundness

If $V[\mathcal{G}] \models \Pi_{\mathcal{G}}$, the engine outputs an *incremental update* V^{Δ} s.t

$$V[\mathcal{G}] :+: V^{\Delta} \models \Pi_{\mathcal{G} \# \Delta}$$

Incremental Δ -Matching

Compute V^Δ s.t $V[\mathcal{G} :+ \Delta] = V[\mathcal{G}] :+ V^\Delta$ with a *delta program*¹

Idea: distributing deltas over joins and factoring.

¹ Ashish Gupta, et al.: *Maintaining Views Incrementally*. SIGMOD 1993: 157-166

Incremental Δ -Matching

Compute V^Δ s.t $V[\mathcal{G} :+ \Delta] = V[\mathcal{G}] :+ V^\Delta$ with a *delta program*¹

Idea: distributing deltas over joins and factoring.

Delta Program

For $V \leftarrow L_1, \dots, L_n$, a *delta program* is a set of clauses of the form

$$V \leftarrow L_1, \dots, L_{i-1}, L_i^\Delta, L_{i+1}^\nu, \dots, L_n^\nu$$

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Incremental Δ -Matching

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Delta Programs

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$$V \leftarrow \mathbf{L}_1, \dots, \mathbf{L}_{i-1}, L_i^\Delta, L_{i+1}^\nu, \dots, L_n^\nu, \text{ where}$$

$\mathbf{L}_j$ is matched against $\mathcal{G}$ atoms with its symbol.

¹ Ashish Gupta, et al.: *Maintaining Views Incrementally*. SIGMOD 1993: 157-166

Incremental Δ -Matching

Compute V^Δ s.t $V[\mathcal{G} :+ \Delta] = V[\mathcal{G}] :+ V^\Delta$ with a *delta program*¹

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Delta Programs

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$$V \leftarrow L_1, \dots, L_{i-1}, \mathbf{L}_i^\Delta, L_{i+1}^\nu, \dots, L_n^\nu, \text{ where}$$

L_j is matched against $\mathcal{G}$ atoms with its symbol.

$\mathbf{L}_j^\Delta$ is matched against Δ atoms with its symbol.

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Incremental Δ -Matching

Compute V^Δ s.t $V[\mathcal{G} :+ \Delta] = V[\mathcal{G}] :+ V^\Delta$ with a *delta program*¹

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Delta Programs

For $V \leftarrow L_1, \dots, L_n$, a *delta program* is a set of clauses of the form

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L_j is matched against $\mathcal{G}$ atoms with its symbol.

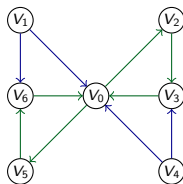
L_j^Δ is matched against Δ atoms with its symbol.

L_j^ν is matched against $\mathcal{G} :+ \Delta$ atoms with its symbol

¹ Ashish Gupta, et al.: *Maintaining Views Incrementally*. SIGMOD 1993: 157-166

Example: Incremental Δ -Matching

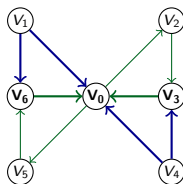
pfriends(X, Y) $\leftarrow$ *knows*(X, Y), *connected*(Z, X), *connected*(Z, Y)



(a) Initial Graph $\mathcal{G}$

Example: Incremental Δ -Matching

pfriends(X, Y) $\leftarrow$ *knows*(X, Y), *connected*(Z, X), *connected*(Z, Y)

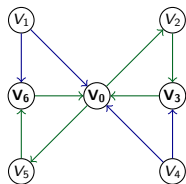


(a) Initial Graph $\mathcal{G}$

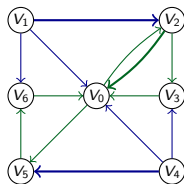
pfriends $[\mathcal{G}] = \{(\mathbf{V}_6, \mathbf{V}_0), (\mathbf{V}_3, \mathbf{V}_0)\}$

Example: Incremental Δ -Matching

$$pfriends(X, Y) \leftarrow knows(X, Y), connected(Z, X), connected(Z, Y)$$



(a) Initial Graph $\mathcal{G}$



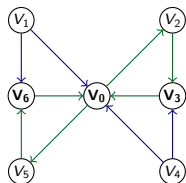
(b) Updated Graph

...updating $\mathcal{G}$ with $\Delta = \{connected(V_1, V_2), knows(V_2, V_0), connected(V_4, V_5)\}$

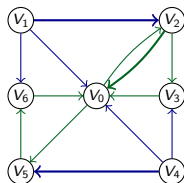
$$pfriends[\mathcal{G} :+ \Delta] = pfriends[\mathcal{G}] \cup ?$$

Example: Incremental Δ -Matching

$$pfriends(X, Y) \leftarrow knows(X, Y), connected(Z, X), connected(Z, Y)$$



(a) Initial Graph $\mathcal{G}$



(b) Updated Graph

...updating $\mathcal{G}$ with $\Delta = \{connected(V_1, V_2), knows(V_2, V_0), connected(V_4, V_5)\}$

$$pfriends[\mathcal{G} :+ \Delta] = pfriends[\mathcal{G}] \cup ?$$

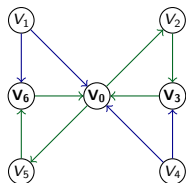
$$pfriends^\Delta(X, Y) \leftarrow knows^\Delta(X, Y), connected^\Delta(Z, X), connected^\Delta(Z, Y)$$

$$pfriends^\Delta(X, Y) \leftarrow knows^\nu(X, Y), connected^\Delta(Z, X), connected^\Delta(Z, Y)$$

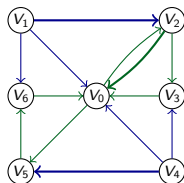
$$pfriends^\Delta(X, Y) \leftarrow knows^\nu(X, Y), connected^\nu(Z, X), connected^\Delta(Z, Y)$$

Example: Incremental Δ -Matching

$$pfriends(X, Y) \leftarrow knows(X, Y), connected(Z, X), connected(Z, Y)$$



(a) Initial Graph $\mathcal{G}$



(b) Updated Graph

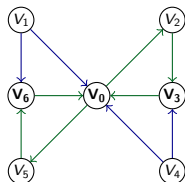
...updating $\mathcal{G}$ with $\Delta = \{connected(V_1, V_2), knows(V_2, V_0), connected(V_4, V_5)\}$

$$pfriends[\mathcal{G} :+ \Delta] = pfriends[\mathcal{G}] \cup \emptyset \cup \dots$$

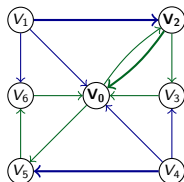
$$pfriends^\Delta(X, Y) \leftarrow knows^\Delta(X, Y), connected(Z, X), connected(Z, Y)$$

Example: Incremental Δ -Matching

$$\textcolor{red}{pfriends}(X, Y) \leftarrow \textcolor{green}{knows}(X, Y), \textcolor{blue}{connected}(Z, X), \textcolor{blue}{connected}(Z, Y)$$



(a) Initial Graph $\mathcal{G}$



(b) Updated Graph

...updating $\mathcal{G}$ with $\Delta = \{\textcolor{blue}{connected}(V_1, V_2), \textcolor{green}{knows}(V_2, V_0), \textcolor{blue}{connected}(V_4, V_5)\}$

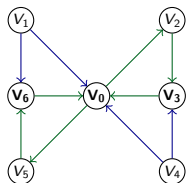
$$\textcolor{red}{pfriends}[\mathcal{G} \text{ :+ } \Delta] = \textcolor{red}{pfriends}[\mathcal{G}] \cup \{(V_2, V_0)\} \cup \dots$$

$$\textcolor{red}{pfriends}^\Delta(X, Y) \leftarrow \textcolor{green}{knows}^\Delta(X, Y), \textcolor{blue}{connected}(Z, X), \textcolor{blue}{connected}(Z, Y)$$

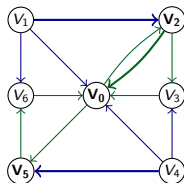
$$\textcolor{red}{pfriends}^\Delta(X, Y) \leftarrow \textcolor{green}{knows}^\nu(X, Y), \textcolor{blue}{connected}^\Delta(Z, X), \textcolor{blue}{connected}(Z, Y)$$

Example: Incremental Δ -Matching

$$pfriends(X, Y) \leftarrow knows(X, Y), connected(Z, X), connected(Z, Y)$$



(a) Initial Graph $\mathcal{G}$



(b) Updated Graph

...updating $\mathcal{G}$ with $\Delta = \{connected(V_1, V_2), knows(V_2, V_0), connected(V_4, V_5)\}$

$$pfriends[\mathcal{G} \vdash \Delta] = pfriends[\mathcal{G}] \cup \{(V_2, V_0)\} \cup \{(V_0, V_2), (V_0, V_5)\}$$

$$pfriends^\Delta(X, Y) \leftarrow knows^\Delta(X, Y), connected^\Delta(Z, X), connected^\Delta(Z, Y)$$

$$pfriends^\Delta(X, Y) \leftarrow knows^\nu(X, Y), connected^\Delta(Z, X), connected^\Delta(Z, Y)$$

$$pfriends^\Delta(X, Y) \leftarrow knows^\nu(X, Y), connected^\nu(Z, X), connected^\Delta(Z, Y)$$

Main Results

- *certified graph query evaluation & maintenance engine*
 - 1062 loc (definitions) + 734 loc (proofs)
 - extracted OCaml engine tested on realistic graph databases
- *machine-checked proofs* of foundational database results
 - mathematical representation of core engine components

Angela Bonifati, **Stefania Dumbrava**, Emilio Jesus Gallego Arias
Certified Graph View Maintenance with Regular Datalog.
Th. and Practice of Logic Programming, 18(3-4):372–389, 2018.

<https://github.com/VerDILog/>

Perspectives

- extension to more expressive graph models
- support for Datalog extensions
- handling complex transactions
- refinement methodology to formally test commercial engines

Thank you !

Related Work

Datalog Setting:

- **Certified Standard and Stratified Datalog Engines**
[Dumbrava, PhD Thesis 2016], [Benzaken et al., ITP 2017]

Graph Setting:

- A Mechanized Formalization of GraphQL [Diaz et al., 2020]
- Incremental Graph Computation for RPQ [Fan et al., 2017]

Relational Setting:

- Certifying SQL Semantics [Chu et al. 2017], [Benzaken et al.. 2019]
- Verified Relational Algebra Query Compilers [Auerbach et al., 2017]
- Verified Relational Data Model [Benzaken et al., 2014]

Experiments

Goal: confirm extracted engine's IVM runtime $<$ its FVM runtime

Setting:

- gMark synthetic datasets and query workloads:
 - WD, the Waterloo SPARQL Diversity Test Suite (Wat-Div)
 - SNB, the LDBC Social Network Benchmark
- schema size: $|supp(\mathcal{G})| = 82$ (WD), $|supp(\mathcal{G})| = 27$ (SNB)
- instance & workload sizes: $|\mathcal{G}| = 1K$, $|\mathcal{W}| = 10$ UC2RPQ
- $\rho_{supp} = \frac{|supp(\Delta^+)|}{|supp(\mathcal{G})|} \in \{0.05, 0.1, 0.15, 0.2, 0.25\}$
- $\rho = \frac{|\Delta^+|}{|\mathcal{G}'|} * 100$
- Time Gain = FVM - IVM, Ratio Gain = $100 - \frac{100 * IVM}{FVM}$

Experiments

ρ_{supp}	ρ	FVM	IVM	Time Gain	Ratio Gain
0.05	1.4%	558.7	484.75	73.95	13.23%
0.1	3.67%	561.89	472.7	89.19	15.87%
0.15	17.93%	562.67	475.96	86.71	15.41%
0.2	9.7%	562.13	476.4	85.73	15.25%
0.25	18.26%	563.4	482.64	80.76	14.33%

Table: $\mathcal{W}_{WD}$ Runtimes (ms) for Varying Support Update Size (ρ_{supp})

ρ_{supp}	ρ	FVM	IVM	Time Gain	Ratio Gain
0.05	10.89%	18.75	10.88	7.87	41.97%
0.1	19.3%	17.77	10.55	7.22	40.63%
0.15	10.77%	17.55	11.68	5.82	33.25%
0.2	26.09%	17.17	11.71	5.46	31.79%
0.25	28.34%	14.71	11	3.71	25.22%

Table: $\mathcal{W}_{SNB}$ Runtimes (ms) for Varying Support Update Size (ρ_{supp})

Experiments - Insights

- *absolute time gain (ms)* of running IVM vs. FVM:
always > 0
- *relative ratio gain (%)* is always better for sparser graphs
SNB runtimes (less dense) $\leq$ WD runtimes (very dense)
- engine works best on bulk updates with small support size
symbol-level maintenance granularity

Incremental Δ -Matching

Let $V \equiv r \bowtie s$, where $V(X, Y) \leftarrow r(X, Z), s(Z, Y)$, r^Δ and s^Δ .

$$V^\Delta = (r^\Delta \bowtie s) \cup (r \bowtie s^\Delta) \cup (r^\Delta \bowtie s^\Delta).$$

$$V^\Delta = (r^\Delta \bowtie s) \cup (r^\nu \bowtie s^\Delta), \text{ where } r^\nu = r \cup r^\Delta.$$

$$V^\Delta = V_1^\Delta \cup V_2^\Delta, \text{ where:}$$

$$\begin{aligned} V_1^\Delta &\leftarrow r^\Delta(X, Z), s(Z, Y) \\ V_2^\Delta &\leftarrow r^\nu(X, Z), s^\Delta(Z, Y). \end{aligned}$$

Regular Datalog in Coq

Variables ($V \ \Sigma : \text{finType}$).

Inductive L := $\square \mid +$.

Inductive egraph := EGraph **of** {set $V * V$ }.

Inductive lrel := LRel **of** {ffun $\Sigma * L \rightarrow \text{egraph}$ }

Record atom := Atom { syma : Σ ; arga : $T * T$ }.

Record lit := Lit { tagl : L; atoml : atom }.

Record cbody := CBody { litb : seq lit }.

Record clause := Clause { headc : $T * T$; bodyc : seq cbody }.

Inductive program := Program **of** {ffun $\Sigma \rightarrow \text{clause } T \ \Sigma \ L$ }.

Incremental View Maintenance

Incremental Atom Matching

$$M_{\mathcal{G},\Delta}^{A,m}(a) = (\text{if } m \in \{\mathbf{B}, \mathbf{F}\} \text{ then } M_{\mathcal{G}}^A(a) \text{ else } \emptyset) \cup (\text{if } m \in \{\mathbf{D}, \mathbf{F}\} \text{ then } M_{\Delta}^A(a) \text{ else } \emptyset)$$

Incremental Body Matching

For a set of body literals $B \equiv [L_1, \dots, L_n]$, generates B_{Δ} :

$$\begin{bmatrix} L_1^{\mathbf{D}} & L_2^{\mathbf{F}} & \dots & L_{n-1}^{\mathbf{F}} & L_n^{\mathbf{F}} \\ L_1^{\mathbf{B}} & L_2^{\mathbf{D}} & \dots & L_{n-1}^{\mathbf{F}} & L_n^{\mathbf{F}} \\ \dots & \dots & \dots & \dots & \dots \\ L_1^{\mathbf{B}} & L_2^{\mathbf{B}} & \dots & L_{n-1}^{\mathbf{B}} & L_n^{\mathbf{D}} \end{bmatrix}$$

Incremental Clausal Maintenance Operator

$$T_{\mathcal{G},\text{supp}}^{\Pi,s}(\Delta) = \begin{cases} T^{\Pi,s}(\mathcal{G} :+ : \Delta), & (s \notin \text{supp}) \vee (\Delta_- \cup D) \neq \emptyset \\ \bigcup_{B_m \in B_{\Delta}} M_{\mathcal{G},\Delta}^B(B_m), & \text{otherwise} \end{cases}$$

Graph Query Complexity

<i>Query Fragment</i>	<i>Evaluation</i>	<i>Containment</i>
RPQ	NLOGSPACE-complete	PSPACE-complete
RPQ_C	$\#P$ -complete	Undefined
2RPQ	NLOGSPACE-complete	PSPACE-complete
C2RPQ	NLOGSPACE-complete	EXPSPACE-complete
UC2RPQ	NLOGSPACE-complete	EXPSPACE-complete
nUC2RPQ	NLOGSPACE-complete	EXPSPACE-complete
UCN2RPQs	NLOGSPACE-complete	EXPSPACE-complete
RQ	NLOGSPACE-complete	2EXPSPACE-complete

Table: Graph query valuation and containment data complexity.

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