



Answering Why-questions by Exemplars in Attributed Graphs

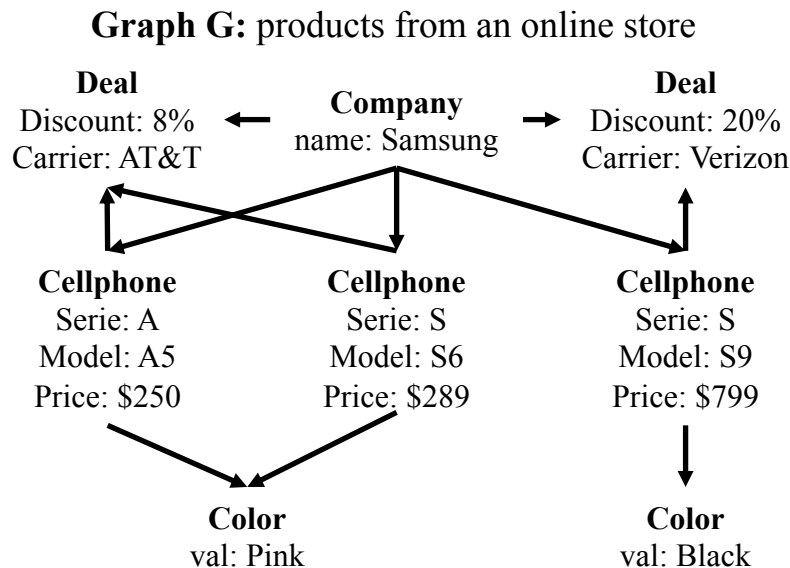
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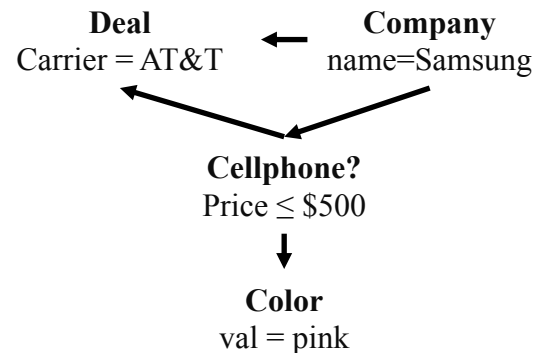
Multi-attributed Graph and Subgraph Query



- Multi-attributed graphs: a labeled graph with attributes on nodes
- Subgraph query Q: a (labeled) graph pattern with a focus node u_o
- Knowledge extraction, POI recommendation, Medical Analysis, etc..



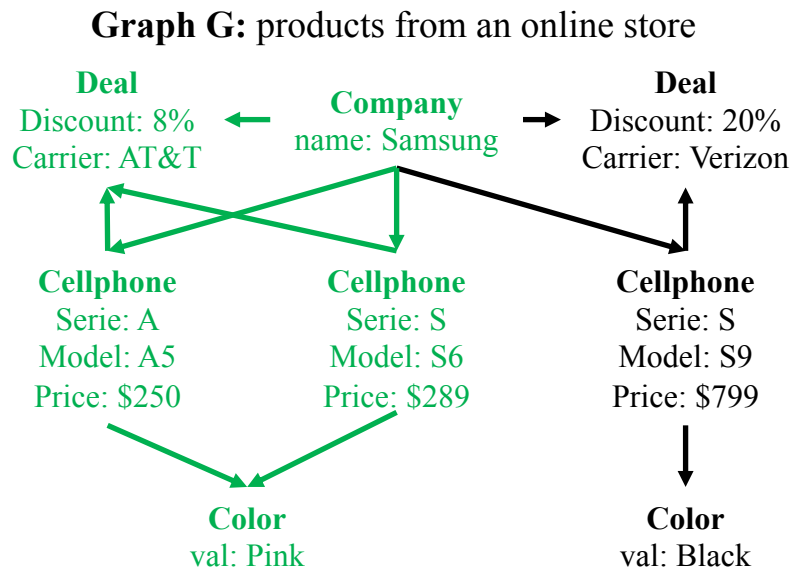
Query Q: Find Samsung cellphones packed with color pink and carrier AT&T, cheaper than \$500



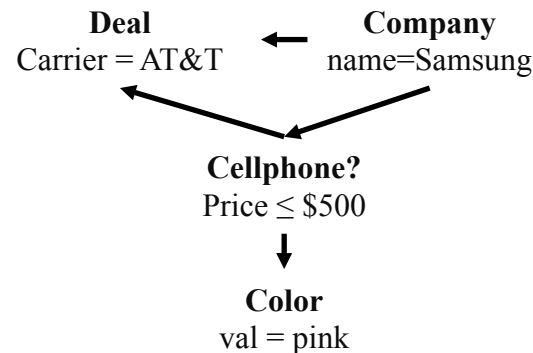
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 **Query Q:** Find Samsung cellphones packed with color pink and carrier AT&T, cheaper than \$500



Matches: {A5, S6}

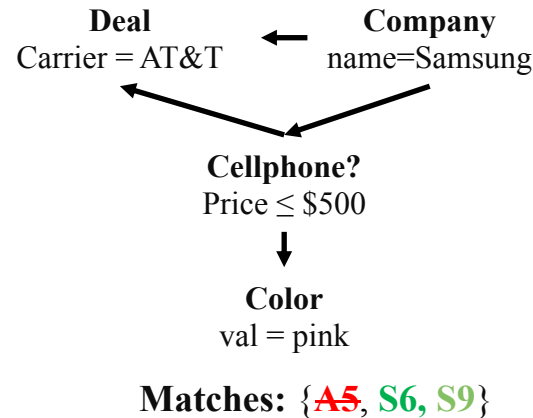
Why-questions by Exemplars

- Why-questions
 - **Why question:** “why some (unexpected) entities are in the query answer?”; and
 - **Why-not question:** “why certain entities are missing from the query result?”
- Exemplars: User specified entities / SQL statement to characterize answers



“I want to see some cellphones which are similar to S9”

Relevant entities: {S6, S9}, Irrelevant entities: {A5}

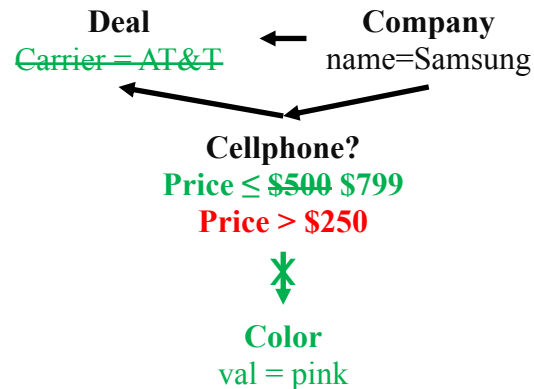
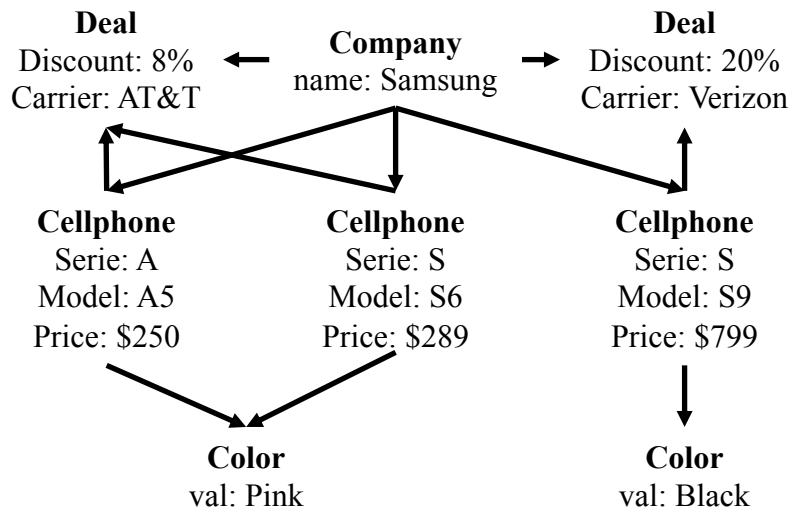


Why-questions by Exemplars: an Example



"I want to see some cellphones which are similar to S9"

Graph G: products from an online store

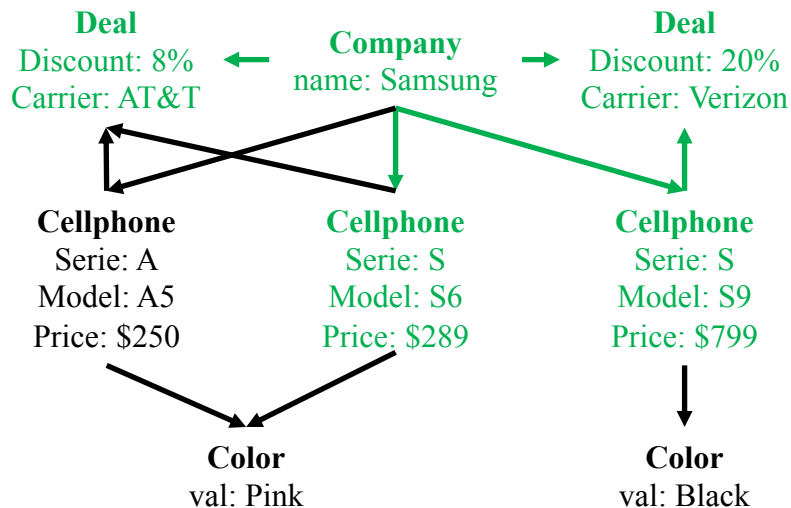


Why-questions by Exemplars: an Example

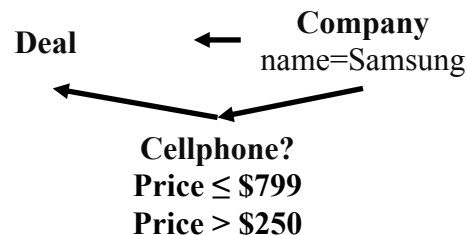


"I want to see some cellphones which are similar to S9"

Graph G: products from an online store



Query rewrite Q'



Matches: {S6, S9}

Editing Operator	Matches
Carrier = AT&T Price ≤ \$799 Cell. - Color	S9
Price > \$250	A5

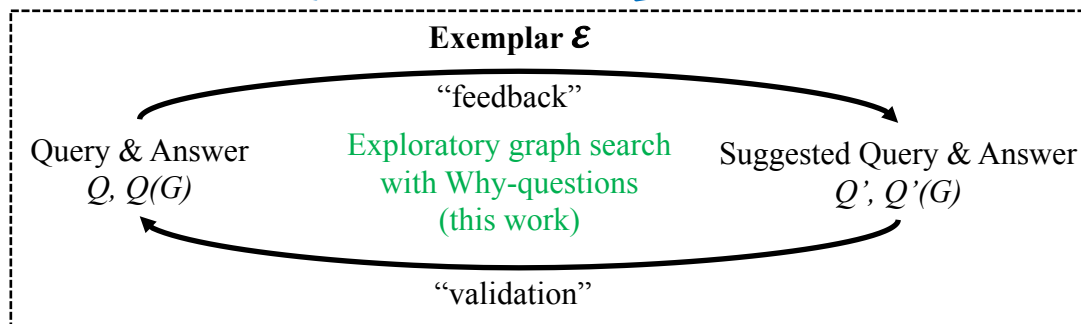
Explanation

Graph Exploratory Search Guided by Examples



A new formalization of answering
Why-questions by exemplars.

Q-Chase: a new symbolic
formalization.



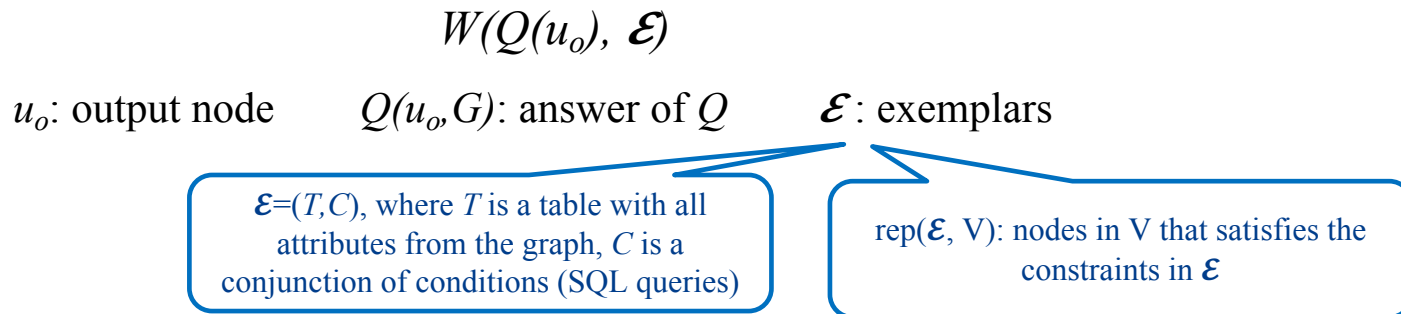
Feasible algorithms with
performance guarantees.

Application (demo): NAVIGATE

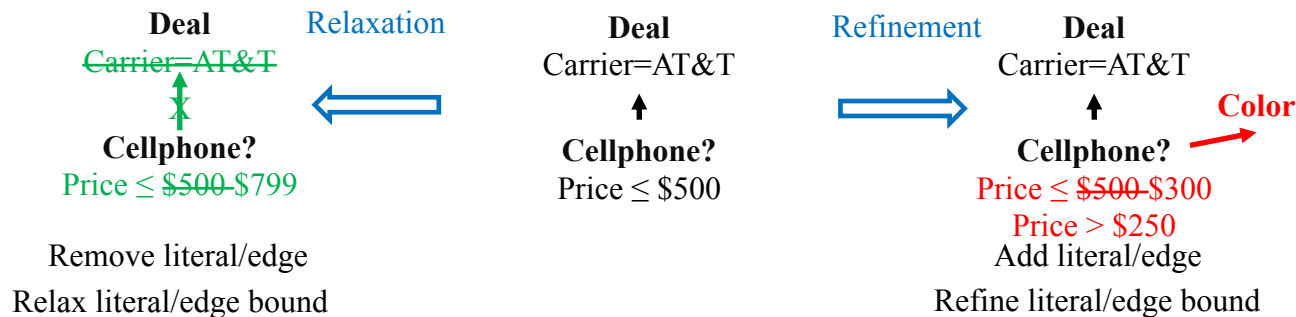
Answering Why-questions: Problem Formulation



- Why-question:



- Answer of Why-questions: query rewrites (operators with editing cost)

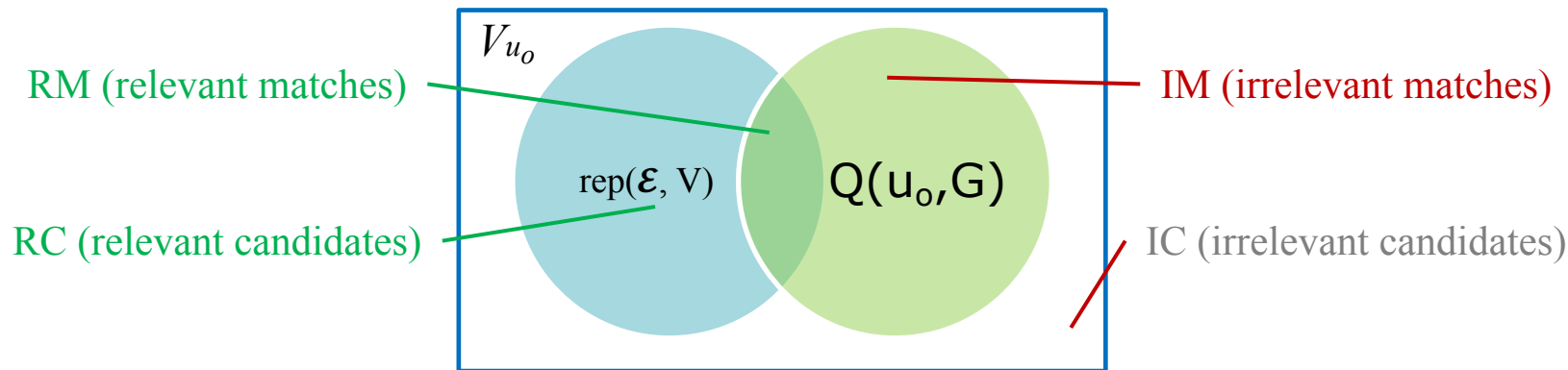




- Answer closeness:

$$cl(v, \mathcal{E}) = \frac{\sum_{v \in RM(\mathcal{E}, Q)} cl(v, \mathcal{E}) - \lambda |IM(\mathcal{E}, Q)|}{|V_{u_o}|}$$

Rewarded and penalized by weighted number of RM and the number of IM in the answer set



Answering Why-questions: Problem Statement



- **Input:** a query Q , answer $Q(G)$, graph G , a Why-question W , and editing budget B ,
- **Output:** a query rewrite $Q' = Q \oplus O^*$, such that

$$O^* = \arg \max_{O: c(O) \leq B} cl(Q'(G), \mathcal{E})$$

- The problem of answering Why question by exemplar is **NP-hard** and **hard to approximate**.
- Solution overview:

Problem	Property	Algorithm	Description
Why-Question	NP-Hard Not in APX	AnsW	Exact
		AnsHeu	Heuristic
Why-Many	Approximable	ApxWhyM	$1/2(1 - 1/e)$ approximation
Why-Empty	P-Time for special case	AnsWE	Exact for special case

A Characterization by Extending Chase



- Q-Chase enforces value constraints from $\mathcal{E}$ over subgraph queries and their answers in G

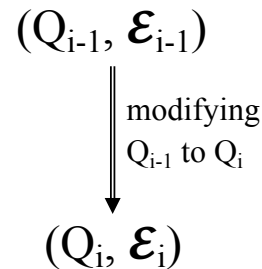
In contract to conventional Chase

- An inductive process

- Initialization: $(Q_0, \mathcal{E}_0) = (Q, \mathcal{E})$
- Q-Chase step i: $(Q_i, \mathcal{E}_i) = \text{Q-Chase}(Q_{i-1}, \mathcal{E}_{i-1})$

- A Q-Chase is terminated when all the constraints in $\mathcal{E}$ are satisfied

- Finding the optimal terminal sequence that maximizes answer closeness.



Computing Optimal Query Rewrite - Algorithm

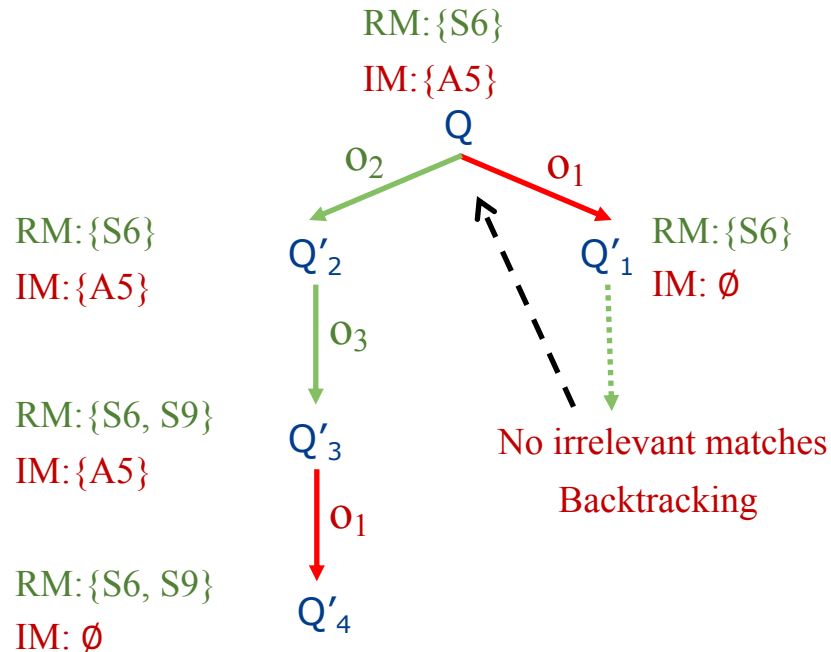


- Our idea: performing a best first search strategy with backtracking to simulate Q-Chase
 - Generate and prioritize a set of “picky” operators; operators that are likely to improve closeness
 - Query construction and evaluation;
 - Terminate when no budget or achieve optimal Q’.

- Performance analysis:
 - Anytime behavior
 - Delay time to simulate a Q-Chase:

$$O(|Q.S(G)|_{b_m}^{|Q|+1})$$

$|Q.S(G)|_{b_m}$: b_m -hop neighbors of the edges in star views in G b_m : maximum edge bound

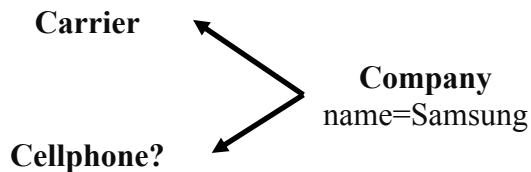


Computing Optimal Query Rewrite - Optimization



- Pruning: closeness upper bound
- Star views: a primitive structure
 - Query decomposition
 - Picky operator suggestion
 - Incremental query evaluation

Star query Q_{11}



Price ≥ 790 RAM ≥ 4
Display ≤ 6.4 Storage ≥ 64

RM = {P₃, P₄, P₅}
IM = {P₁, P₂}

Star table T_{11}

Company	Cell-Carrier	Cell-Company
P ₁	AT&T	Samsung
P ₂	T-Mobile	Samsung
P ₃	Sprint	Samsung
P ₄	Sprint	Samsung
P ₅	Sprint	Samsung

AddL(Carrier.name = 'Sprint')

AddL(Carrier.name = 'AT&T')

Picky



- Datasets

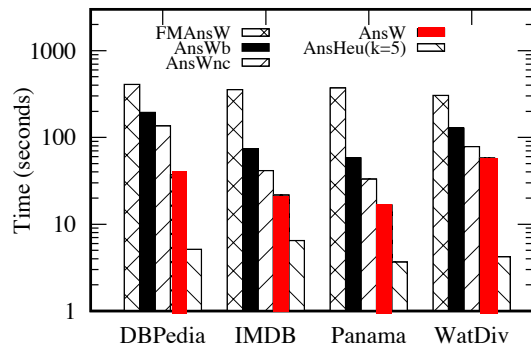
Name	Description	# of nodes	# of edges	# of attributes per node
DBPedia	Knowledge Graph	4.86M	15M	9
IMDb	Movie Network	1.7M	5.2M	6
Offshore	Financial Activities	839K	3.6M	4
WatDiv	e-commerce information (synthetic)	521K	9.1M	8

- Ground truth queries: benchmark queries and a query generator.
- Question generation: disturb $(Q^*, Q^*(G))$ by injecting random atomic operators.
- Algorithms
 - **Exact algorithms:** AnsW (optimized), AnsWnc (no caching), AnsWb (no pruning).
 - **Heuristics:** AnsHeu, and AnsHeuB (random).
 - **Baseline:** FMAAnsW (Mottin et. al. 2015).

Experimental Result

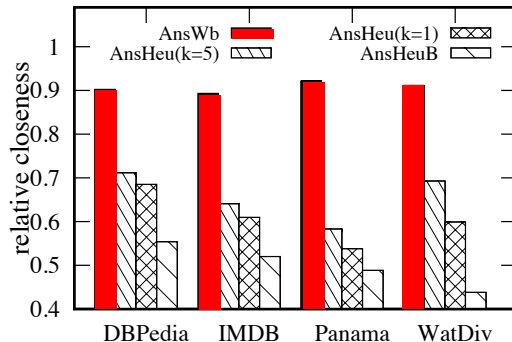


■ Answering Why questions: efficiency and effectiveness



AnsW outperforms FMAnsW, AnsWb, and AnsWnc by 10.82, 3.41, and 2.1 times, respectively

AnsW constantly achieves the maximum closeness among others.

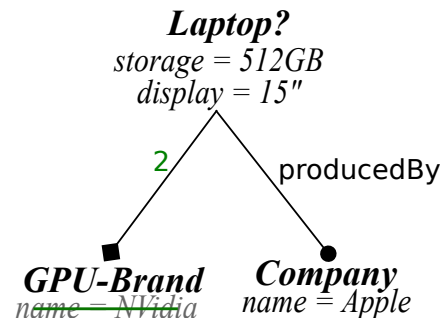


■ User study:

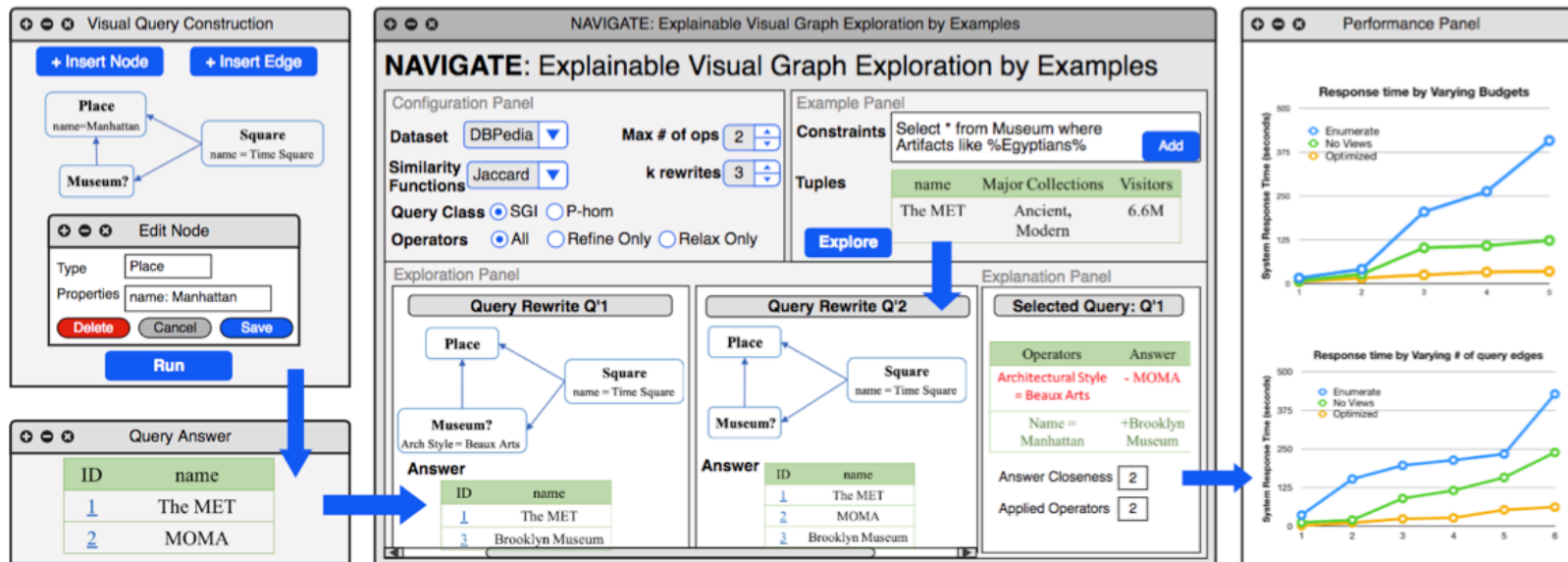
- Ask users to re-rank the top-3 query rewrites from AnsW;
- Good answer relevance: $nDCG_3 = 0.71$.

■ Case analysis: product recommendation

- A user searched for recent computer models with GPU (after 2018)
- Desired laptops powered by either Intel or AMD GPU



A Demo system



NAVIGATE: Explainable Visual Graph Exploration by Examples

Wednesday & Thursday 16:20 – 17:50



Thank you!

