

# Preferentially Annotated Regular Path Queries

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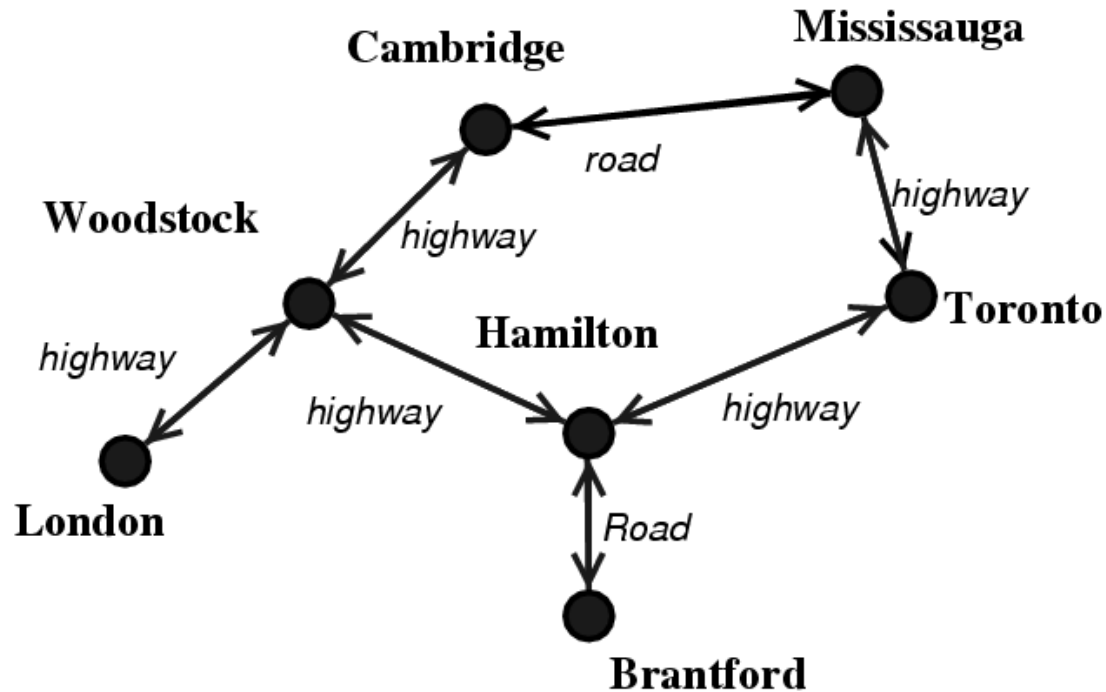
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# Regular Path Queries (RPQ's)

- Essentially regular expressions over database labels.



- E.g.:  $Q = \text{highway}^*$
- Meaning: Find highway routes.

# RPQ's vs. Datalog

Semantically RPQ's are a fragment of Datalog.

However,

- They are easier for people to use
  - they have used reg.ex. from the early days of computers
- Important reasoning services on RPQ's are decidable
  - E.g. Containment/Equivalence is decidable for RPQ's, while for Datalog it's not.

# Not any database path, but...

- Surely, I prefer highways, but can tolerate one road:

$$Q = \text{highway}^* \cdot \text{road} \cdot \text{highway}^*$$

- Well, I prefer highways, but can tolerate up to  $k$  roads or city streets:

$$Q = \text{highway}^* \parallel (\text{road} + \text{street} + \varepsilon)^k$$

# Preferences: Boolean Way

$$Q = \text{highway}^* \parallel (\text{road} + \text{street} + \varepsilon)^k$$

- Pair of objects will be produced as an answer if there exists a path between them satisfying the user query.
- There is just a “yes” or “no” qualification for the query answers.
- But, answers aren’t equally good!
  - A pair of objects connected by a **highway path with only 1 intervening road** is a “better” answer than a pair of objects connected by a **highway path with 5 intervening roads**.

# A simple syntactic addition

- User can annotate the symbols in the regular expressions with “markers” (typically natural numbers), which “strengthen” or “weaken” his (pattern) preferences.

$$Q = (\text{highway:0})^* \parallel (\text{road:1} + \text{street:2} + \varepsilon)^k$$

- Meaning:
  - User ideally prefers **highways**,
  - then **roads**, which he prefers less,
  - and finally he can tolerate **streets**, but with an even lesser preference.

# Semantics (Quantitative)

$$Q = (\text{highway:0})^* \parallel (\text{road:1} + \text{street:2} + \varepsilon)^k$$

- The system should produce:
  - first the pairs of objects connected by highways,
  - then the pairs of objects connected by highways intervened by 1 road,
  - and so on.
- The “so on” raises some important semantical questions.
- Is a pair of objects connected by a **highway path intervened by two roads** equally good as another pair of objects connected by a **highway path intervened by one street only**?
- Indeed, in this example, it might make sense to consider them equally good, and “concatenate” weights by **summing them up**.

# Qualitative Semantics

$$Q = (\text{viarail:0})^* \parallel (\text{greyhound:1} + \text{aircanada:2} + \epsilon)^k.$$

- Is now a pair of objects connected by  
a path with two **greyhound** segments  
equally preferable as a pair of objects connected  
with one **aircanada** segment?
- If the user is **afraid of flying**, she might want to “concatenate”  
edge-weights by choosing the **maximum** of the weights.
- Then an itinerary with no matter how many **greyhound** segments  
is preferable to an itinerary containing only one **flight** segment.



# Hybrid Semantics

$$Q = (\text{viarail:0})^* \parallel (\text{greyhound:1} + \text{aircanada:2} + \varepsilon)^k.$$

- Following a purely **qualitative** approach,  
**greyhound** itineraries  
are always preferable to  
**itineraries containing aircanada segments**,  
while these itineraries are equally preferable, no matter how many lags the flight has.
- Sometimes we need to distinguish among itineraries on the same “**level of discomfort**.”
  - Namely, we should be able to (quantitatively) say for example that  
**a direct aircanada route**  
is preferable to  
**an aircanada route with a stop-over**,  
which again is preferable to  
**an aircanada route with three lags**.

# Semirings

- In total, from all the above, we have four kind of preference semantics:
  - Boolean,
  - quantitative,
  - qualitative,
  - hybrid.
- In all these semantics, we:
  - aggregate (“concatenate”) preference weights along edges of the paths, and then
  - aggregate path preferences when there are multiple paths connecting a pair of objects.
- We regard the preference annotations as elements of a semiring, with two operations:
  - “plus”
  - “times”
- The “times” aggregates the preferences along edges of a path, while the “plus” aggregates preferences among paths.

# Semirings

$\mathcal{R} = (\mathbf{R}, \oplus, \otimes, \mathbf{0}, \mathbf{1})$  such that

$(\mathbf{R}, \oplus, \mathbf{0})$  is a commutative monoid with  $\mathbf{0}$  as the identity element for  $\oplus$ .

$(\mathbf{R}, \otimes, \mathbf{1})$  is a monoid with  $\mathbf{1}$  as the identity element for  $\otimes$ .

$\otimes$  distributes over  $\oplus$ : for all  $x, y, z \in \mathbf{R}$ ,

$$(x \oplus y) \otimes z = (x \otimes z) \oplus (y \otimes z)$$

$$z \otimes (x \oplus y) = (z \otimes x) \oplus (z \otimes y).$$

Natural order  $\leq$  on  $\mathbf{R}$ :  $x \leq y$  iff  $x \oplus y = x$

# Annotated Queries

$\mathcal{R} = (\mathbf{R}, \oplus, \otimes, \mathbf{0}, \mathbf{1})$  semiring.

An  $\mathcal{R}$ -annotated query  $Q$  over  $\Delta$  is a function

$$Q : \Delta^* \rightarrow \mathbf{R}.$$

We write  $(w, x) \in Q$  instead of  $Q(w) = x$ .

- When annotated queries are given by “**annotated regular expressions**,” we have annotated regular path queries (**ARPQ**'s).

# Annotated Automata

- Computationally, ARPQ's are represented by “annotated automata”  $(P, \Delta, \mathcal{R}, \tau, p_0, F)$
- The language defined by an annotated automaton A is:

$$[A] = \{(w, x) \in \Delta^* \times \mathbf{R} :$$

$$w = r_1 r_2 \dots r_n,$$

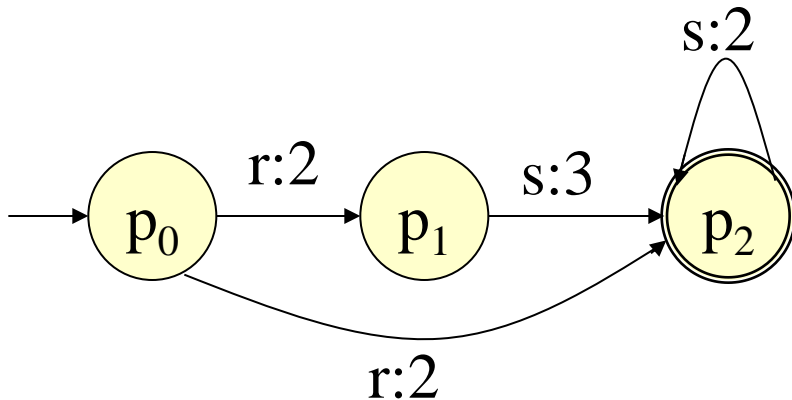
$$x = \oplus \{x_1 \otimes \dots \otimes x_n :$$

$$(p_0, r_1, x_1, p_1) \in \tau,$$

...

$$(p_{n-1}, r_n, x_n, p_n) \in \tau,$$

$$p_n \in F\} \}.$$



$$(rs, 4) \in [A]$$

# Query Answers

Given a database DB, and  
an annotated Q over semiring  $\mathcal{R} = (\mathbf{R}, \oplus, \otimes, \mathbf{0}, \mathbf{1})$

$$\text{Ans}(Q, \text{DB}, \mathcal{R}) = \{(a, b, x) : \\ x = \oplus \{y : (w, y) \in Q \text{ and} \\ w \text{ labels some path from } a \text{ to } b \text{ in DB}\}.$$

We have  $(a, b, \mathbf{0})$  as an answer to Q, if there is no path in DB spelling some word in Q.

# Preference Semirings

Boolean preferences:  $\mathfrak{B} = (\{T, F\}, \vee, \wedge, F, T)$

Quantitative preferences:  $\mathfrak{Q} = (\mathbb{N} \cup \{\infty\}, \min, +, \infty, 0)$

Qualitative preferences:  $\mathfrak{F} = (\mathbb{N} \cup \{\infty\}, \min, \max, \infty, 0)$

# Preference Semirings

Hybrid preferences:  $\mathfrak{K} = (\mathbf{R}, \oplus, \otimes, \mathbf{0}, \mathbf{1})$

- Interface again is  $\mathbf{N}$ .
- However,  $\mathbf{R}$  is bigger to allow for a finer ranking

$$\mathbf{R} = \{0, 1, 1^{(2)}, \dots, 2, 2^{(2)}, \dots\} \cup \{\infty\}$$

1, 2, ... are shorthand for  $1^{(1)}, 2^{(1)}, \dots$

- $n^{(i)}$  :  $n$  -- level of discomfort,  
 $i$  -- how many times we are “forced to endure”  
that level of discomfort.

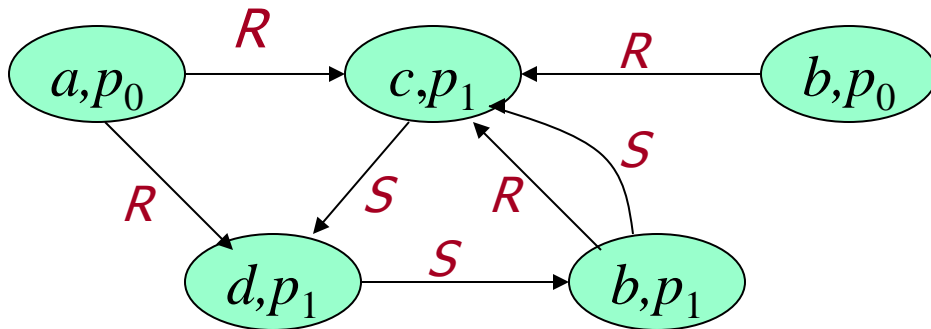
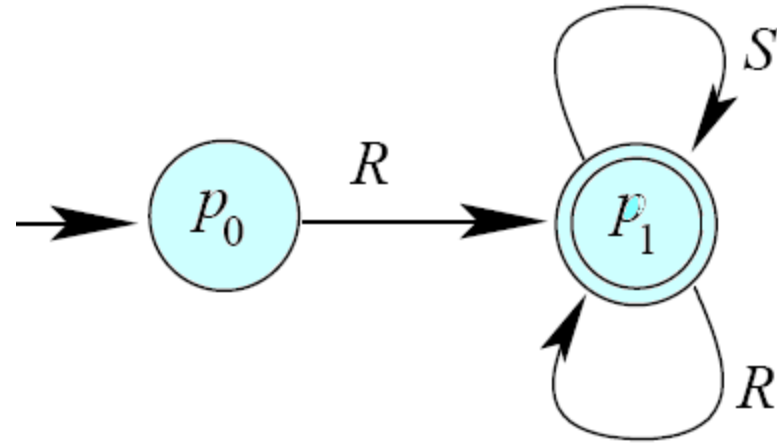
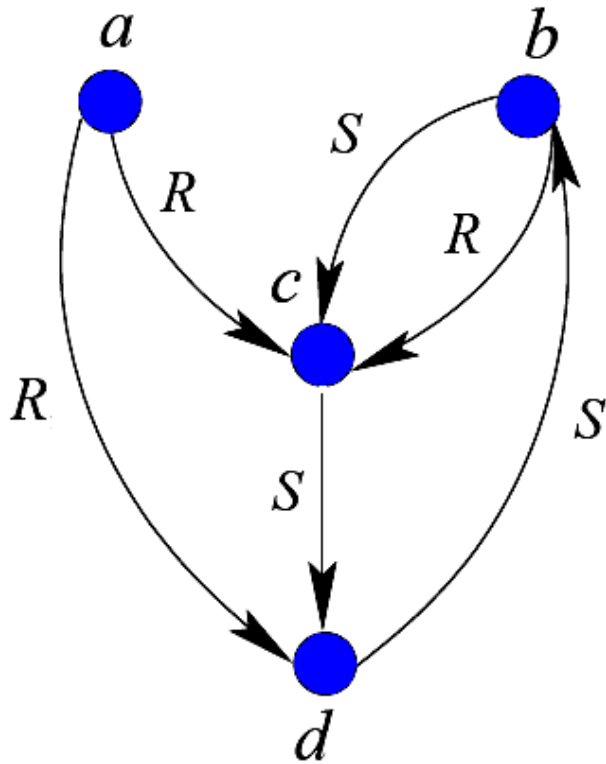
$$n^{(i)} \oplus m^{(j)} = \begin{cases} n^{(i)} & \text{if } n < m \\ m^{(j)} & \text{if } n > m \\ n^{(\min\{i,j\})} & \text{if } n = m, \end{cases} \quad n^{(i)} \otimes m^{(j)} = \begin{cases} n^{(i)} & \text{if } n > m \\ m^{(j)} & \text{if } n < m \\ n^{(i+j)} & \text{if } n = m \end{cases}$$



# Hybrid Preferences

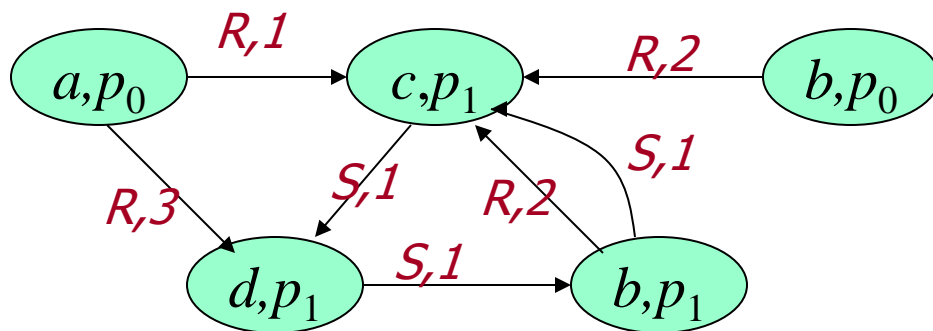
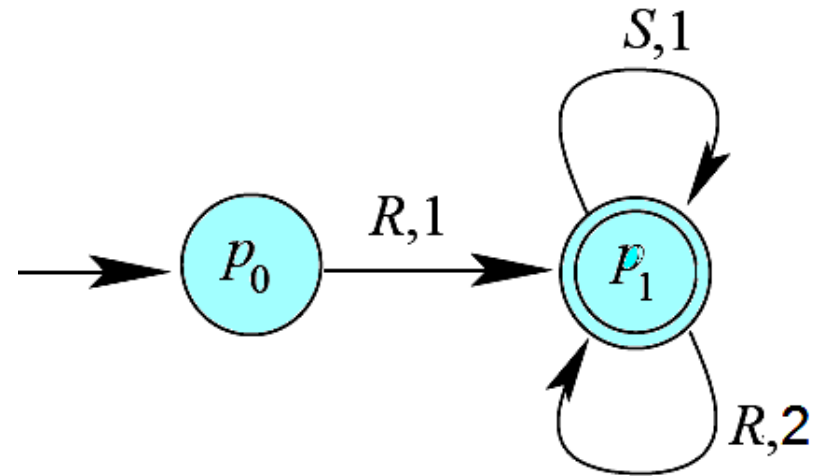
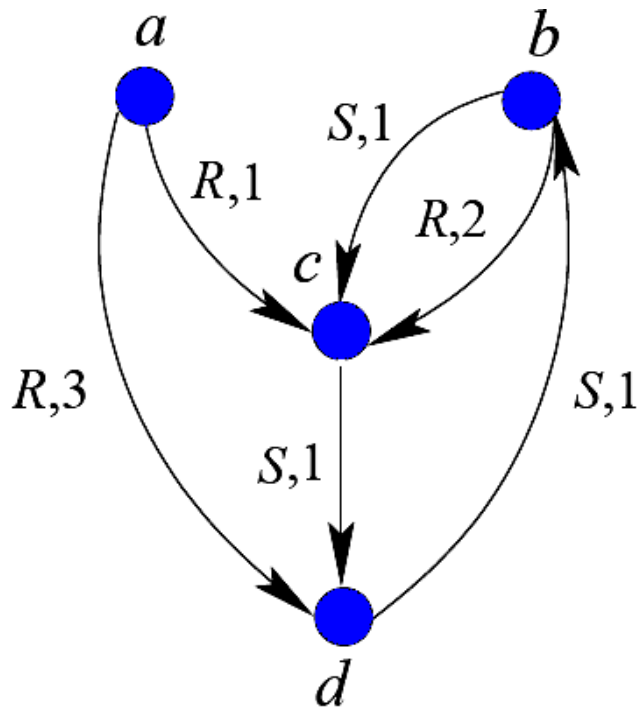
- The user, annotates query symbols with natural numbers representing his preferences.
- Similarly with **qualitative semantics**, only database edges matched by transitions annotated with the “worst” **level of discomfort** will really count.
- Similarly with **quantitative semantics**, paths with same “**worst-level of discomfort**” are comparable.
  - Namely, the best path will be the one with the **fewest “worst-level of discomfort”** edges.

# Answering of RPQ's (Classical)



Then, do **reachability** in the green graph.

# Answering of ARPQ's



Then, compute **generalized shortest paths** in the **green graph**.

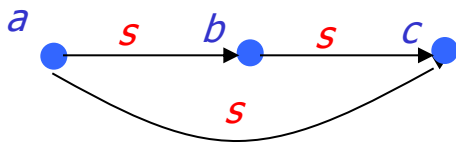
# LAV Data Integration

- No database in the classical sense.
- We have “data-sources,” characterized by a definition over a “global schema”:  $\Delta = \{R, \dots\}$
- Each data-source also has a name, and the set of these names constitutes the “local schema”:  $\Omega = \{s_1, \dots, s_n\}$
- Mapping:  $\text{def}(s_i) = S_i$
- LAV system also has a set of tuples over the local schema.
- Queries are formulated on the global schema.
- Data exists in the local schema, so, a translation from  $\Delta$  to  $\Omega$  has to be performed in order to be able to compute query answers.

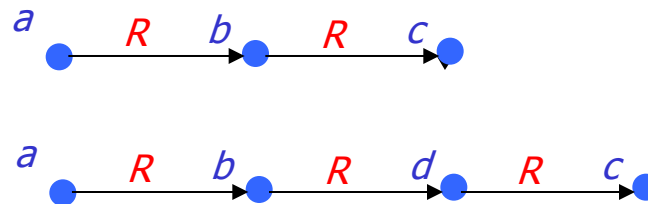
# Source Collections and Possible DB's

- Let  $\Omega = \{s_1, \dots, s_n\}$  be the local schema.
- Then, a source collection  $\mathfrak{S}$  is a graph database on  $\Omega$ .
- $poss(\mathfrak{S})$ : Set of all databases from which the given source collection  $\mathfrak{S}$  might have been generated.
- E.g., consider  $S=R^*$  and

$\mathfrak{S}$



Possible DB's



...

Under **sound**  
source  
assumption

# Certain Answer

$$\text{CAAns}(Q, \mathfrak{S}) = \bigcap_{\text{DB} \in \text{poss}(\mathfrak{S})} \text{Ans}(Q, \text{DB})$$

How to express this using the Boolean Semiring?

$$\text{CAAns}(Q, \mathfrak{S}, \mathfrak{B}) = \bigwedge_{\text{DB} \in \text{poss}(\mathfrak{S})} \text{Ans}(Q, \text{DB}, \mathfrak{B})$$

where

$$\text{Ans}(Q, \text{DB}_1, \mathfrak{B}) \wedge \text{Ans}(Q, \text{DB}_2, \mathfrak{B}) =$$

$$\{(a, b, x \wedge y) :$$

$$(a, b, x) \in \text{Ans}(Q, \text{DB}_1, \mathfrak{B}) \text{ and}$$

$$(a, b, y) \in \text{Ans}(Q, \text{DB}_2, \mathfrak{B})\}$$

# Dual Operator and Certain Answer

- We aggregated the answers on possible DB's by using  $\wedge$ , which is the dual of  $\vee$ , which is the  $\oplus$  of  $\mathfrak{B}$  semiring.
- Generalizing, we define

$$x \odot y = \begin{cases} x & \text{if } x \oplus y = y \\ y & \text{if } x \oplus y = x \end{cases}$$

$$\text{CAAns}(Q, \mathfrak{S}, \mathfrak{R}) = \bigodot_{\text{DB} \in \text{poss}(\mathfrak{S})} \text{Ans}(Q, \text{DB}, \mathfrak{R})$$

# Differently said...

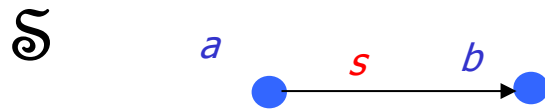
- A tuple  $(a, b, x) \in \text{CAAns}(Q, \mathcal{S}, \mathcal{R})$ , with  $x \neq \mathbf{0}$ ,  
iff  
for each  $\text{DB} \in \text{poss}(\mathcal{S})$  there exists  $y \leq x$  s.t.  
 $(a, b, y) \in \text{Ans}(Q, \text{DB}, \mathcal{R})$ .
- Definition reflects:  
**certainty** that objects  $a$  and  $b$  are always connected  
with paths, which are preferentially weighted not  
more than  $x$ .



# Practically

Query:  $Q = (\text{highway} : 0)^* \parallel (\text{road} : 1 + \varepsilon)^*$

Source Collection:



Source Definition:  $S = \text{highway}^* \parallel (\text{road} + \varepsilon)^5$

Possible Databases:

All those, which have at least a path (between  $a$  and  $b$ ) labeled by **highways** intervened by at most 5 **roads**.

# Quantitative Semiring

- $\odot$  is *max*, and we have  $(a, b, 5)$  as a certain answer.
- Weight of **5** states our **certainty** that in any possible database, there is a path from *a* to *b*, whose preferential weight w.r.t. the given query is not more than **5**.
- Also, there exists a possible database in which the best path between a and b is exactly **5**.

# Qualitative Semiring

- $\odot$  is again *max*, but we have  $(a, b, 1)$  as a certain answer.
- Weight of **1** states our certainty that in any possible database, there is a path from *a* to *b*, and the **level of discomfort** (w.r.t. the query) for traversing that path is not more than **1**.

# Hybrid Semiring

$$n^{(i)} \odot m^{(j)} = \begin{cases} m^{(j)} & \text{if } n < m \\ n^{(i)} & \text{if } n > m \\ n^{(\max\{i,j\})} & \text{if } n = m \end{cases}$$

- We have  $(a, b, 1^{(5)})$  as a certain answer.
- Because although the **level of discomfort** of the best path connecting  $a$  with  $b$  in any possible database is  $1$ , in the worst case (of such best paths), we need to endure up to  $5$  times such discomfort (w.r.t. the query).
- Of course  $1^{(5)}$  is infinitely better than  $2$ .

# Certain Answers via Query Spheres

- Given  $Q$ , the  $y$ -sphere of  $Q$  is

$$Q^y = \{(w, x) \in \Delta^* \times \mathbb{R} : (w, x) \in Q \text{ and } x \leq y\}$$

- Call them “spheres” because:  $Q^x \subseteq Q^y \subseteq Q$  for  $x \leq y$

- Discrete Semirings:

$$\forall x \exists \text{ “the next element” } y$$

i.e.  $x < y$  and there isn't  $z$ , s.t  $x < z < y$

- Theorem.

$$(a, b, y) \in \text{CAAns}(Q, \mathcal{S}, \mathcal{R}) \text{ iff}$$

$$(a, b, T) \in \text{CAAns}(Q^y, \mathcal{S}, \mathcal{B}) \text{ and}$$

$$(a, b, T) \notin \text{CAAns}(Q^x, \mathcal{S}, \mathcal{B})$$

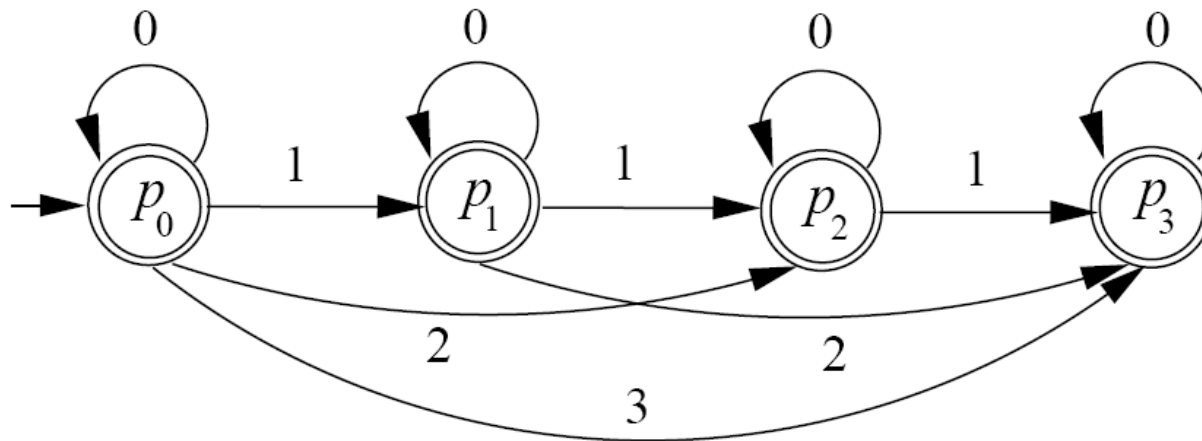
# Certain Answers via Query Spheres

- We know how to compute the certain answer in the Boolean (classical) case: **Calvanese, Di Giacomo, Lenzerini, Vardi, ICDE'00**.
- And, we present next how to compute query spheres.
- **But, is there an upper limit in the index of the spheres?**
- Answer:
  - For the qualitative semiring there is always such bound.
  - For the quantitative and hybrid semirings, we reduce the problem to the **Limitedness Problem** in distance automata introduced and solved by Hashiguchi.
    - If there is such limit, then all the certain answers can be ranked.
    - Otherwise, the certain answers can be computed, but **eventually** ranked.
    - In practice, the user can provide a bound for the quality of certain answers he is interested in.

# Computing Query Spheres

Computing  $Q^k$

- **Qualitative**: Keep only transitions weighted  $\leq k$ .
- **Quantitative**: Intersect with mask automaton (e.g.):

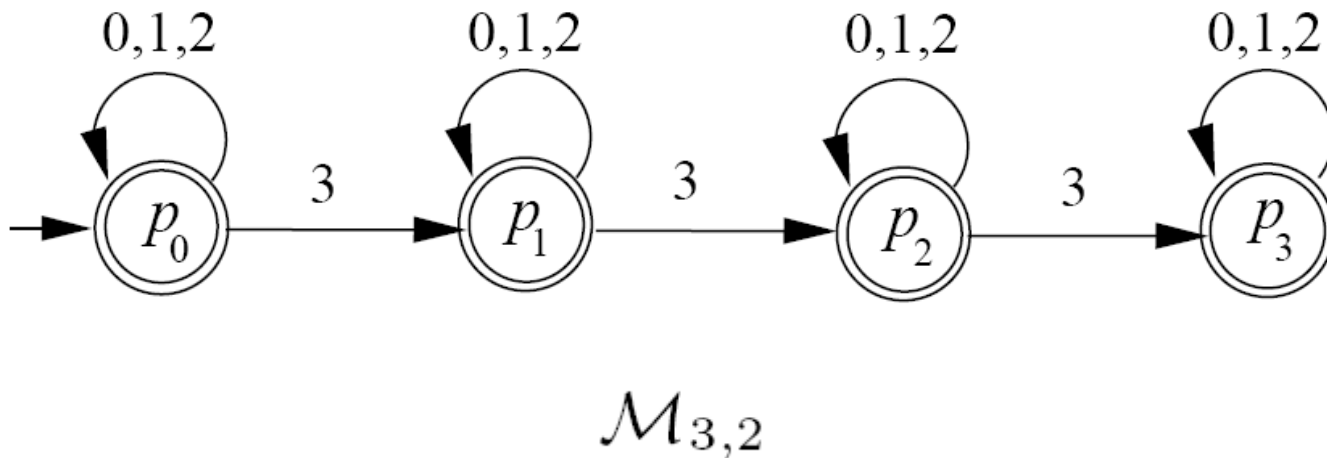


$\mathcal{M}_3$

# Computing Query Spheres: Hybrid

- Computing  $Q^y$  where  $y = n^{(k)}$
- Intersect with a mask automaton, which extracts from the query automaton all the paths with
  - (a) any number of transitions weighted strictly  $< n$ , and
  - (b) not more than  $k$  transitions weighted exactly  $n$ .

E.g.





# Containment and Equivalence (Full Paper)

- We also study in detail the query containment for various semirings.
- We show that the containment is decidable for **deterministic queries**.
- **Allauzen, Mohri, TCS 328, 2004.**  
Show that large classes of weighted NFA's can successfully be determinized.

# Conclusions

- Introduced **preferential regular path queries**
  - whose symbols are annotated with preference weights for “scaling” up or down the intrinsic importance of matching a symbol against a database edge label.
- Different specializations for the same syntactic annotations.
- Various semantics in a unifying semiring framework.
- Studied three important aspects:
  - (1) **query answering**
  - (2) **(certain) query answering in LAV data-integration systems**
  - (3) **query containment and equivalence.**
- In all these, obtained important positive results, which encourage the use of our preference framework for enhanced querying of semistructured databases.

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