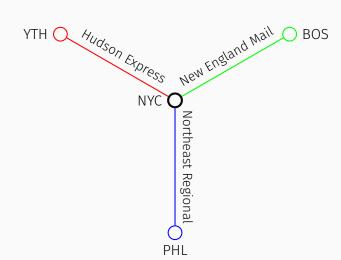
REGULAR PROGRAMMING FOR QUANTITATIVE PROPERTIES OF DATA STREAMS

Rajeev Alur Dana Fisman Mukund Raghothaman IBM PL Day, 2015

University of Pennsylvania



Time		Line	Station
06:00:0) AM	Hudson Express	YTH
06:27:1	3 AM	Northeast Regional	NYC
07:24:20) AM	Hudson Express	YTH
07:28:2	3 AM	Northeast Regional	PHL
08:12:4	6 AM	Hudson Express	YTH
09:08:54	1 AM	Northeast Regional	NYC
09:35:5	AM 6	New England Mail	NYC
09:43:10) AM	New England Mail	BOS
		New England Mail	BOS
10:45:4	L AM	New England Mail	BOS
10:46:2	5 AM	Northeast Regional	PHL
11:32:5	2 AM	Northeast Regional	NYC
11:35:0	AM 6	Northeast Regional	PHL
11:58:0) AM	New England Mail	BOS
12:44:3	7 PM	Northeast Regional	PHL
		Hudson Express	YTH
01:16:3	2 PM	Hudson Express	NYC
		Northeast Regional	PHL
01:49:3	2 PM	Hudson Express	YTH
		Hudson Express	YTH
		Hudson Express	YTH
		Hudson Express	NYC
		Northeast Regional	PHL
		New England Mail	NYC
		New England Mail	BOS
		New England Mail	NYC
		Northeast Regional	NYC
07:08:0	PM	Hudson Express	YTH

THE QUERY

"What is the average travel time from PHL to YTH?"

Time		Line	Station
06:00:0) AM	Hudson Express	YTH
06:27:1	3 AM	Northeast Regional	NYC
07:24:20) AM	Hudson Express	YTH
07:28:2	3 AM	Northeast Regional	PHL
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		Northeast Regional	PHL
		New England Mail	NYC
		New England Mail	BOS
		New England Mail	NYC
		Northeast Regional	NYC
07:08:0	PM	Hudson Express	YTH

	Time		Line	Station
П	06:00:00) AM	Hudson Express	YTH
П	06:27:18	3 AM	Northeast Regional	NYC
	07:24:20) AM	Hudson Express	YTH
ı	07:28:23	3 AM	Northeast Regional	PHL
	08:12:46	AM 6	Hudson Express	YTH
	09:08:54	I AM	Northeast Regional	NYC
Ĭ	09:35:56	MA d	New England Mail	NYC
	09:43:10) AM	New England Mail	BOS
Ц			New England Mail	BOS
Ц			New England Mail	BOS
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Ц			Northeast Regional	PHL
Ц			Hudson Express	YTH
Ц			Hudson Express	NYC
Ц			Northeast Regional	PHL
Ц			Hudson Express	YTH
Ц			Hudson Express	YTH
Ц			Hudson Express	YTH
Ц			Hudson Express	NYC
Ц			Northeast Regional	PHL
Ц			New England Mail	NYC
Ц			New England Mail	BOS
Ц			New England Mail	NYC
Ц			Northeast Regional	NYC
Ц	07:08:09	PM	Hudson Express	YTH

Tin	1e		Line	Station	
			Hudson Express	YTH	
06:	27:18	ΑM	Northeast Regional	NYC	
				YTH	
07:	28:23	ΑM	Northeast Regional	PHL	
08:	12:46	ΑM	Hudson Express	YTH	
09:	:08:54	ΑM	Northeast Regional	NYC	
09:	35:56	ΑM	New England Mail	NYC	
			New England Mail	BOS	
			New England Mail	BOS	
10:	45:41	AM	New England Mail	BOS	
10:	46:25	AM	Northeast Regional	PHL	
			Northeast Regional	NYC	
_			Northeast Regional	PHL	
			New England Mail	BOS	
12:	:44:37	PM	Northeast Regional	PHL	
				YTH	
			Hudson Express	NYC	
			Northeast Regional	PHL	
				YTH	
_				YTH	_
				YTH	
				NYC	
			Northeast Regional	PHL	
			New England Mail	NYC	
			New England Mail	BOS	
			New England Mail	NYC	
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09:43:10	AM	New England Mail	BOS	
10:07:59	AM	New England Mail	BOS	
10:45:41	AM	New England Mail	BOS	
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		Northeast Regional	PHL	
		Hudson Express	YTH	
		Hudson Express	YIH	
		Hudson Express	YTH	
		Hudson Express	NYC	
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		New England Mail	NYC	
		New England Mail	BOS	
		New England Mail	NYC	
		Northeast Regional	NYC	
07:08:09	PM	Hudson Express	YTH	

· Function f_{trip} calculates travel time for a single trip

$$\cdot f_{avg} = iter-avg(f_{trip})$$

· Function f_{trip} calculates travel time for a single trip $f_{trip} = t_{alight} - t_{board}$

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$$t_{board} = split((\neg NER@PHL)^*, NER@PHL \mapsto time, (\neg NER@NYC)^*, NER@NYC, R_{NY})$$

 $\cdot f_{avg} = iter-avg(f_{trip})$

INTRODUCING DREX

· Language based on the idea of function combinators

INTRODUCING DREX

- · Language based on the idea of function combinators
- DReX is a simple, expressive programming model for stream processing, with strong theoretical foundations and fast evaluation algorithms

Basic functions: $data \mapsto cost$

Conditional choice: felse g

Concatenation: split-plus(f, g), split-min(f, g), split-left(f, g),

split-right(f,g), ...

Function iteration: iter-plus(f), iter-max(f), iter-avg(f),

iter-mdn(f), ...

Cost operations: $f_1 + f_2$, $f_1 - f_2$, $\max(f_1, f_2)$, ...

```
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```

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Function iteration: *iter-plus(f), iter-max(f), iter-avg(f),*

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Cost operations: op(f_1, f_2, \dots, f_p)
```

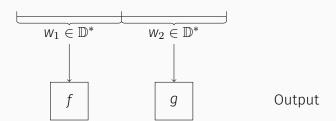
```
Basic functions: data \mapsto cost

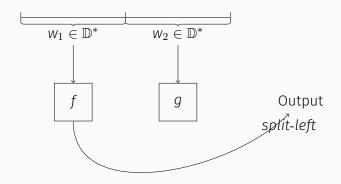
Conditional choice: f else g

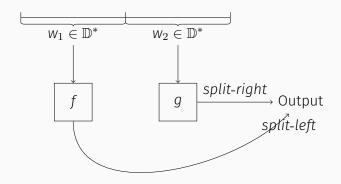
Concatenation: split-plus(f,g), split-min(f,g), split-left(f,g), split-right(f,g), ...

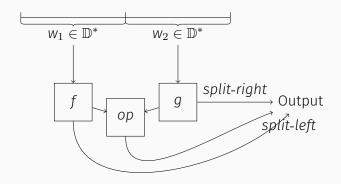
Function iteration: iter-plus(f), iter-max(f), iter-avg(f), iter-mdn(f), ...

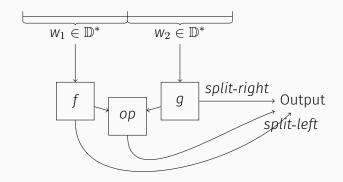
Cost operations: op(f_1, f_2, \dots, f_b)
```



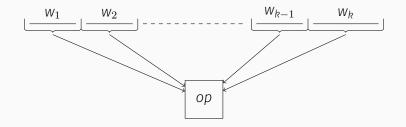


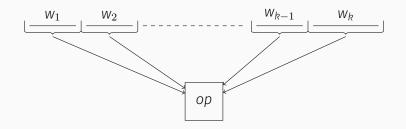




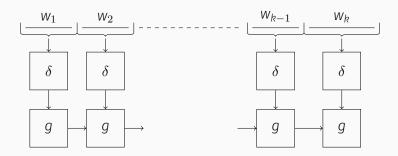


Are these combinators "sufficient"?

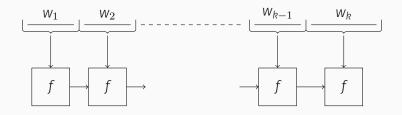




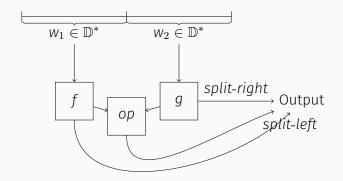
- · Data stream is a sequence of bank transactions
- · Interest applied at end of each month



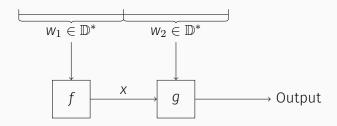
- · Data stream is a sequence of bank transactions
- · Interest applied at end of each month



- · Data stream is a sequence of bank transactions
- · Interest applied at end of each month
- · Map input streams to terms over the cost domain



Are these combinators "sufficient"?



Are these combinators "sufficient"?

```
Basic functions: data \mapsto cost

Conditional choice: f else g

Concatenation: split-plus(f,g), split-min(f,g), split-left(f,g), split-right(f,g), ...

Function iteration: iter-plus(f), iter-max(f), iter-avg(f), iter-mdn(f), ...

Cost operations: op(f_1, f_2, \ldots, f_k)
```

Basic functions: $data \mapsto cost$, $regex \mapsto term$

Conditional choice: felse g

Concatenation: $split(f \rightarrow^{x} g)$, $split(f \leftarrow^{x} g)$

Function iteration: $iter^{\rightarrow}(f)$, $iter^{\leftarrow}(f)$

Cost operations: $op(f_1, f_2, ..., f_k), f[x/g]$

Basic functions: $data \mapsto cost$, $regex \mapsto term$

Conditional choice: felse g

Concatenation: $split(f \rightarrow^{\times} g)$, $split(f \leftarrow^{\times} g)$

Function iteration: $iter^{\rightarrow}(f)$, $iter^{\leftarrow}(f)$

Cost operations: $op(f_1, f_2, ..., f_k), f[x/g]$

Structural operators decoupled from cost operators!

Function descriptions are modular

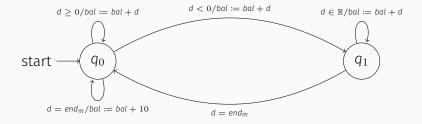
```
currLoc: {phl, nyc, yth, ...}
totalTime: int, tripCount: int
tBoard: int
while exists next event e:
   if e is (ner, phl) and currLoc = phl:
     tBoard := e.time
     currLoc := ...
elif ...:
```

DReX allows regular parsing of the input data stream

- · Unrestricted global choice: felse g
- · Iteration patterns part of query, not of the data: $iter-avg(f_{trip})$

- · Expressively equivalent to regular string-to-term transformations
- Multiple characterizations: MSO-definable graph transformations, streaming string-to-term transducers, two-way finite state transducers
- · Closed under various operations: regular look-ahead, input reversal, function composition etc.

Streaming string-to-term transducers



Taste of the completeness proof

- · Piggy-back on DFA to regular expression translation Construct $R^{i}(q, q')$: all strings from q to q' while only traversing states less than q_{i}
- · Construct $f^i(q, q', x)$: express final value of register x as a DReX expression

Taste of the completeness proof

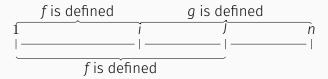
- · Capture data flows using "shapes"
- Construct a partial order over shapes, and use as basis for induction

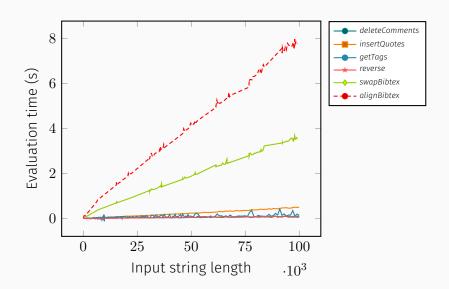


One-pass evaluation algorithm for unambiguous expressions

- f(w) can be computed in time $O(|w| \cdot poly(|f|))$
- · Separating intent from evaluation
- · Space-efficient evaluation where possible
 - *, +, -, min, max, avg
- · Approximation algorithms for the masses: iter-median

- · Require that expressions be "unambiguous"
- · Informally, restricts number of alternative parse trees
- · Not burdensome





STATUS

Done: Expressiveness theorems and evaluation

algorithms

Alur, Freilich, R, LICS 2014

Alur, D'Antoni, R, POPL 2015

Alur, Fisman, R, Unpublished 2015

Ongoing: Approximation algorithms, case studies, static

analysis

THANK YOU!

THANK YOU:

TRY IT OUT AT HTTP://DREXONLINE.COM