

# Synthesis-Powered Visualization Authoring

<https://falx.cs.washington.edu>

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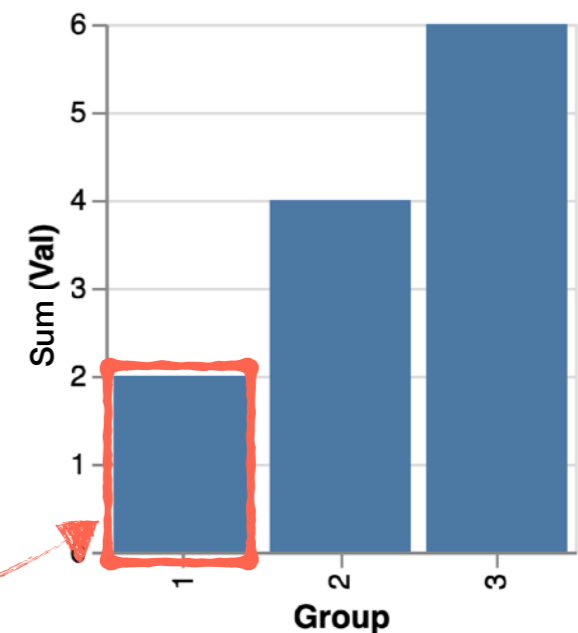
# What is visualization?

“The grammar of graphics”

Visualization = mappings from *data columns* to *geometric properties*

Group	Val
1	2
2	4
3	6

`bar(x ← Group,  
height ← Val)`

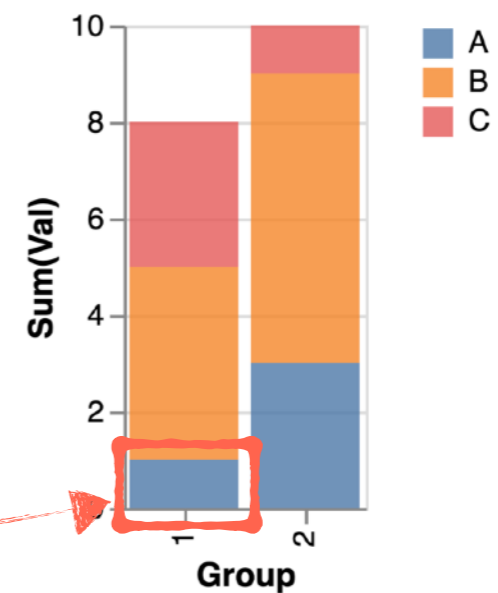


## Visualization in practice

Task: Sum up values for each group in each column and map to a stacked bar chart

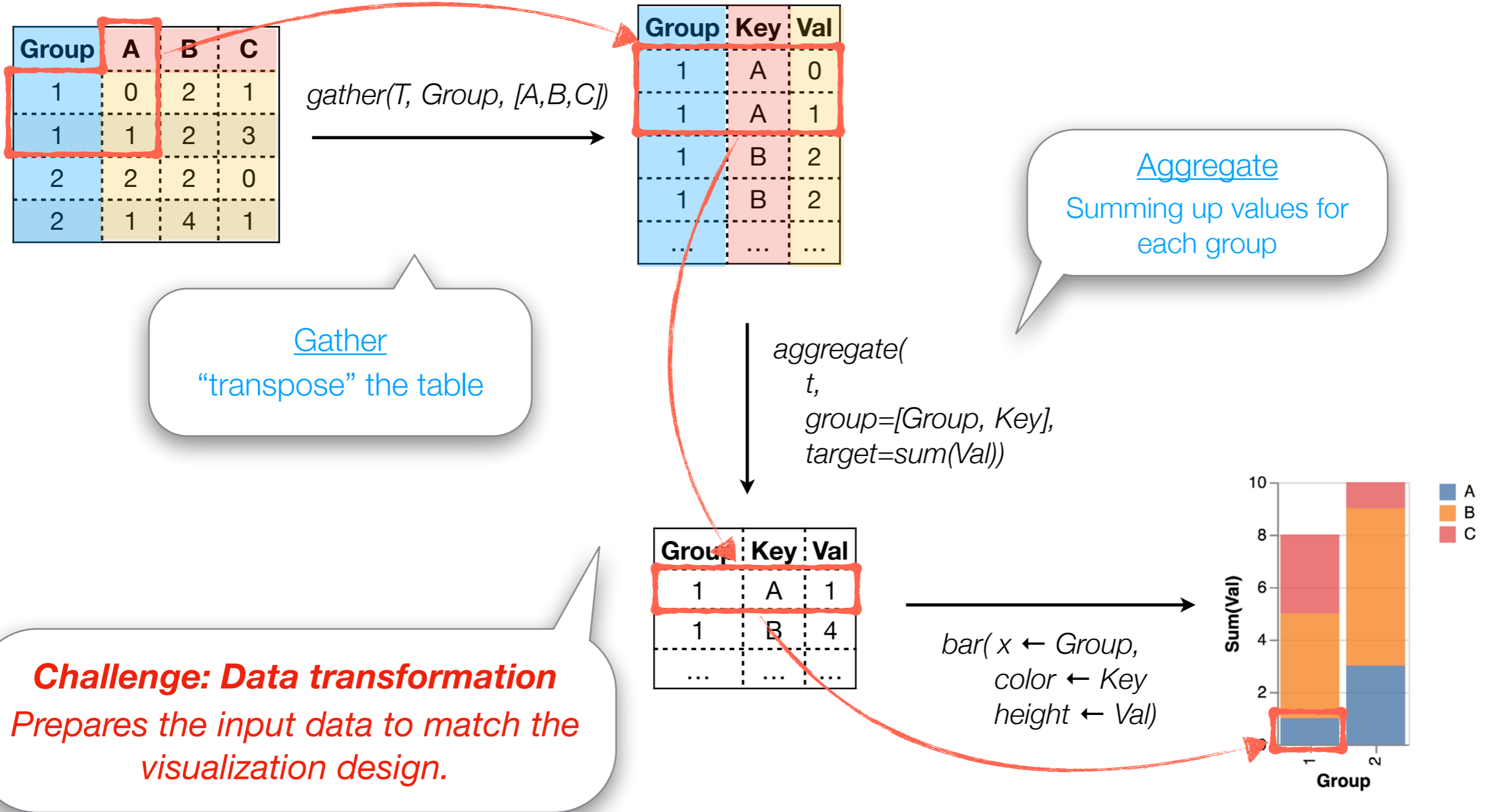
Group	A	B	C
1	0	2	1
1	1	2	3
2	2	2	0
2	1	4	1

?



The visualization expects an input data with three columns  
Each maps to *x*, *height*, *color*

In practice: Visualization = data transformation + design specification

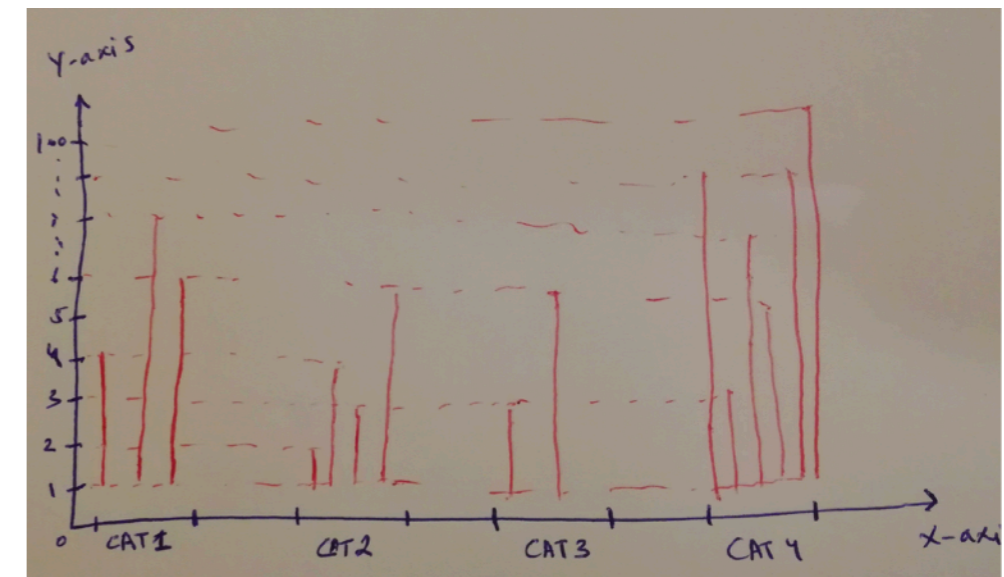
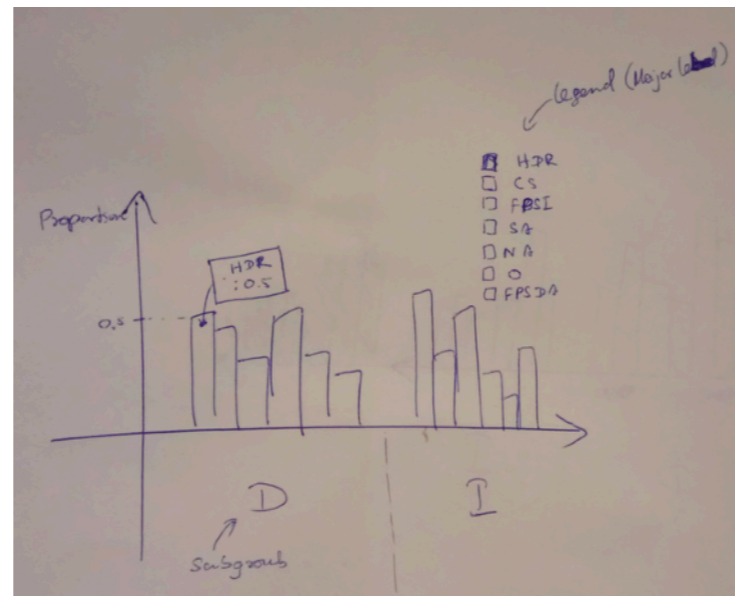
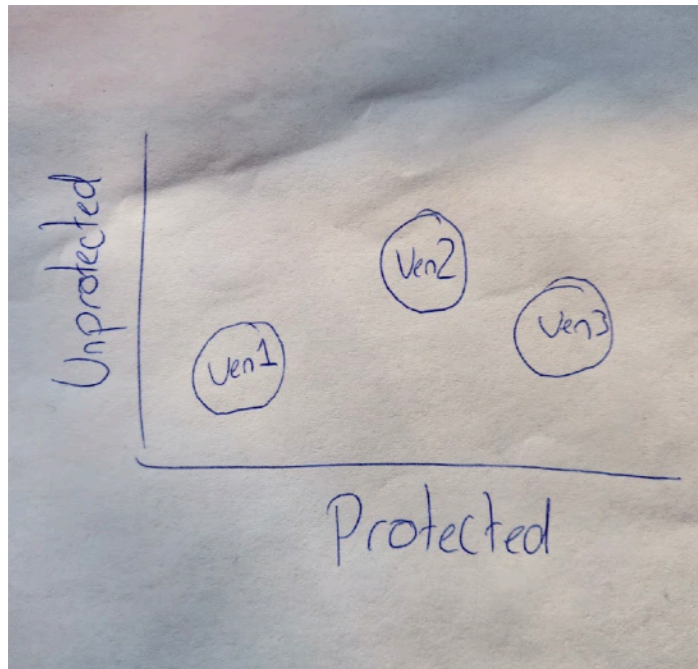
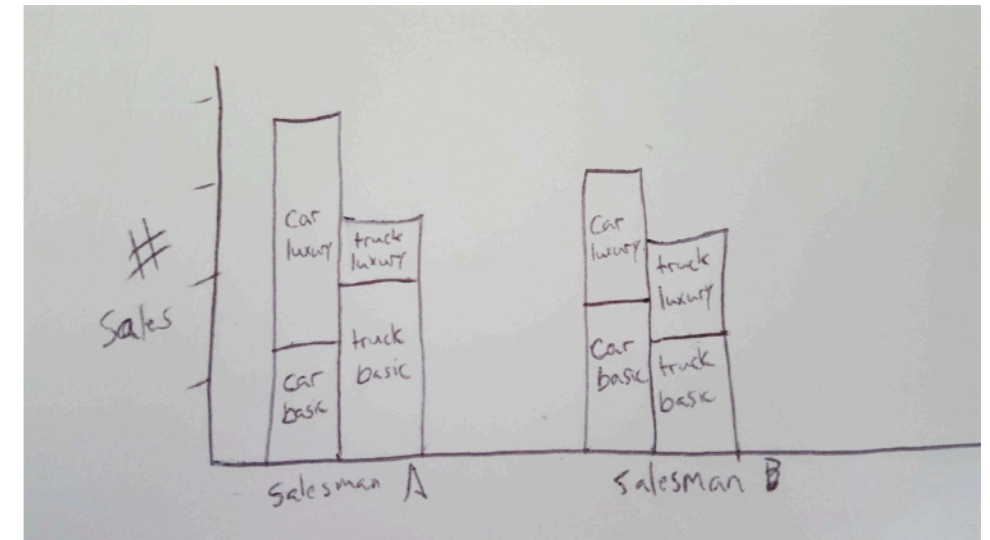
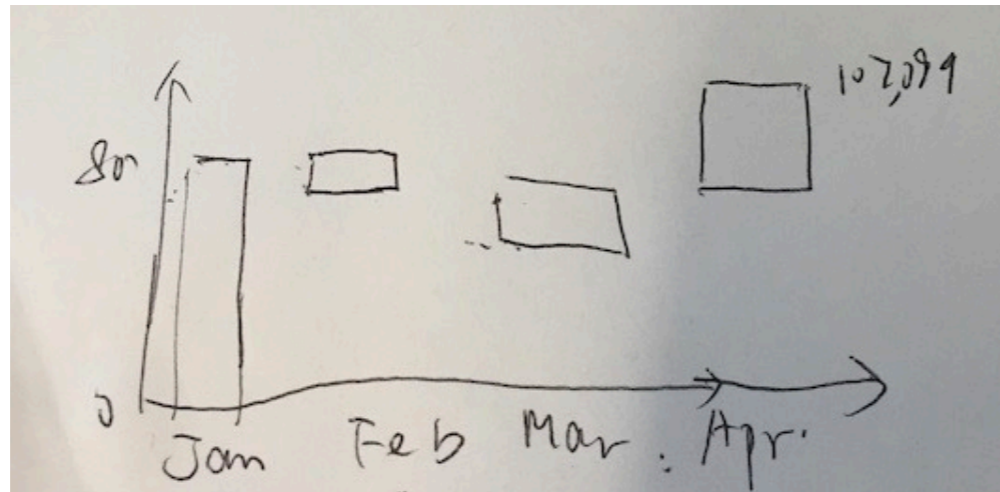


[1] Gatto, Malu AC. "Making research useful: Current challenges and good practices in data visualization." (2015).

[2] Perkel, Jeffrey M. "Data visualization tools drive interactivity and reproducibility in online publishing." *Nature* 554.7690 (2018): 133-134.

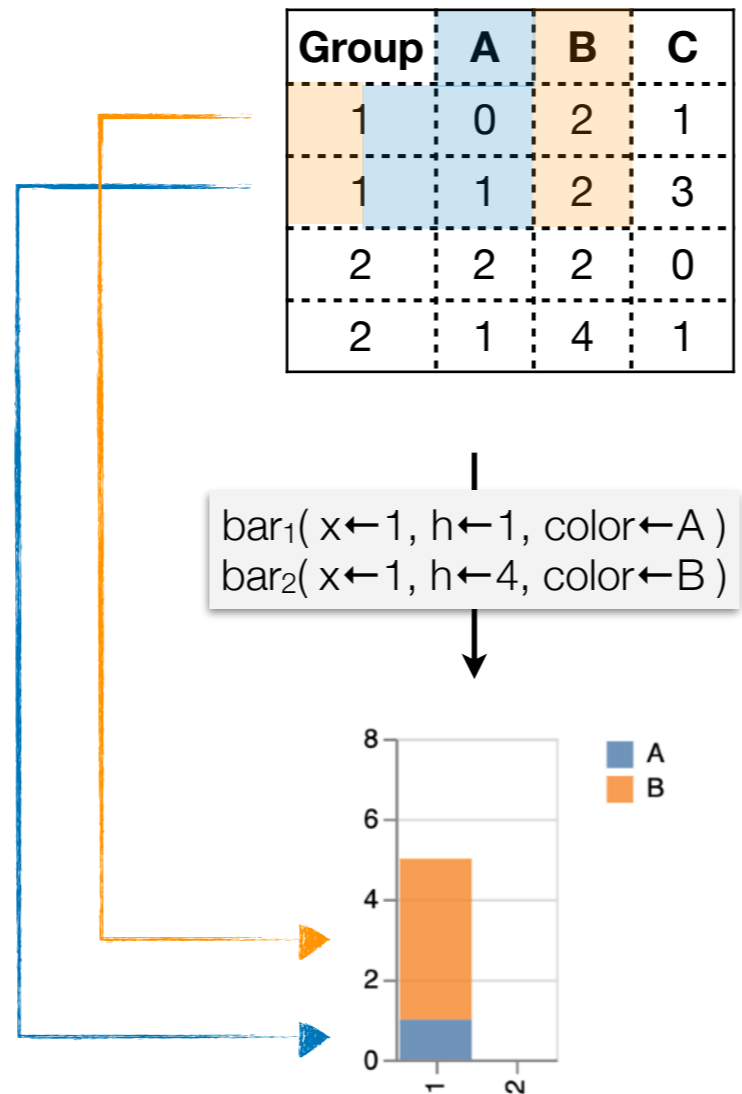
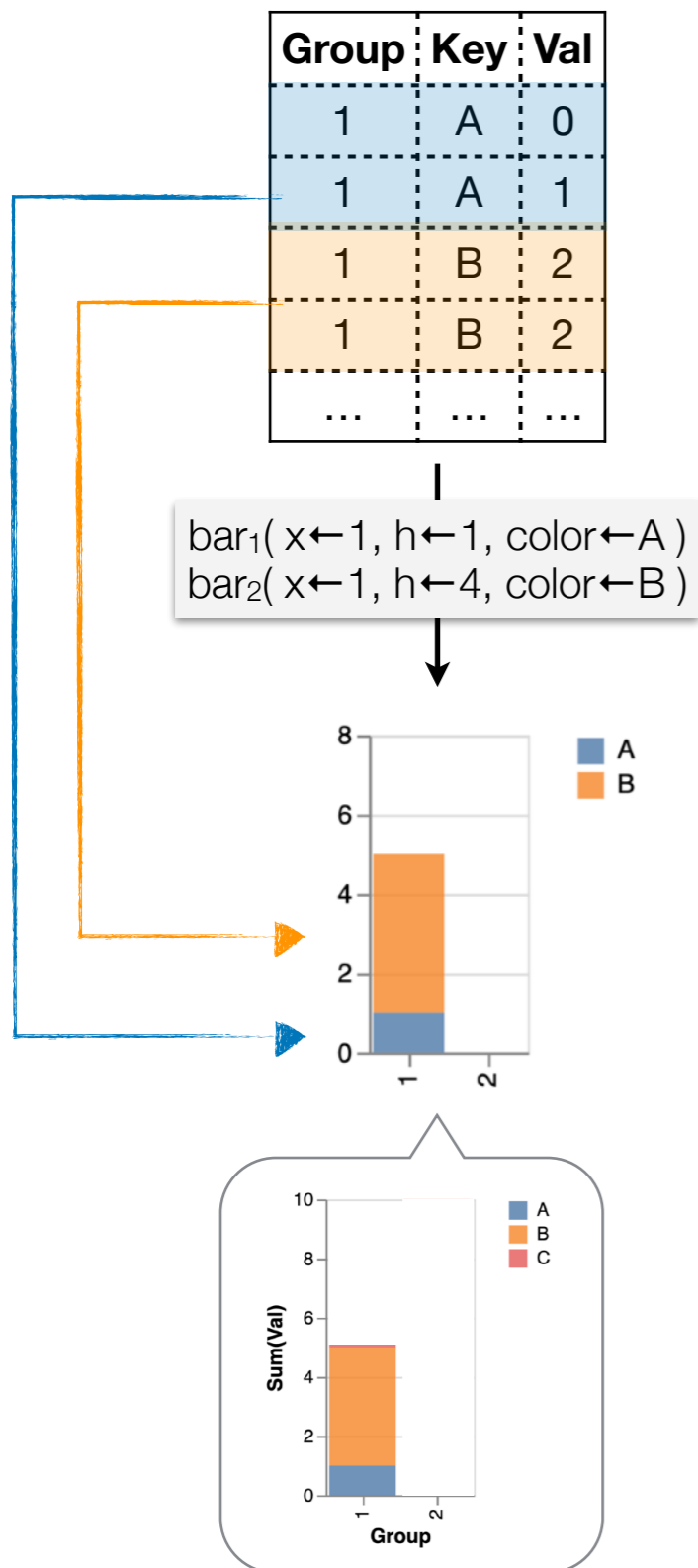
[3] Kasica et al. "Table Scraps: An Actionable Framework for Multi-Table Data Wrangling From An Artifact Study of Computational Journalism" *InfoVis* (2020)

# How end users communicate visualization ideas in practice?



Observation: End users use partial visualizations to communicate their ideas.

# Our design: Visualization by Example



*How can we synthesize visualization from examples?*

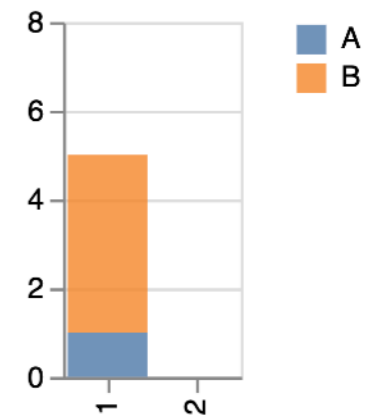
*Easy to specify  
Data-layout independent*

# Synthesis Task

How to synthesize two programs at once?



$V_{partial}$



$T$

Group	A	B	C
1	0	4	3
1	1	3	2
2	3	2	1
2	2	6	1

$p_T$

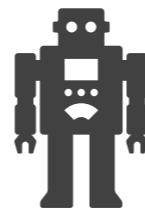
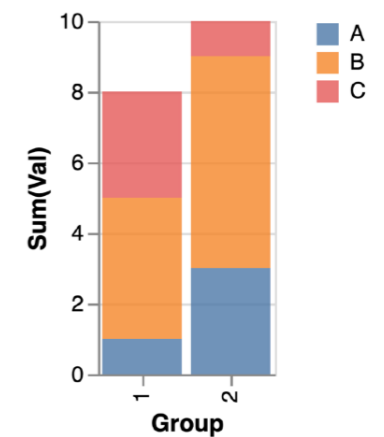


$T'$

$p_V$



$\cup$



Synthesize  $p_T, p_V$  such that the synthesized visualization  $p_V(p_T(T))$  contains  $V_{partial}$



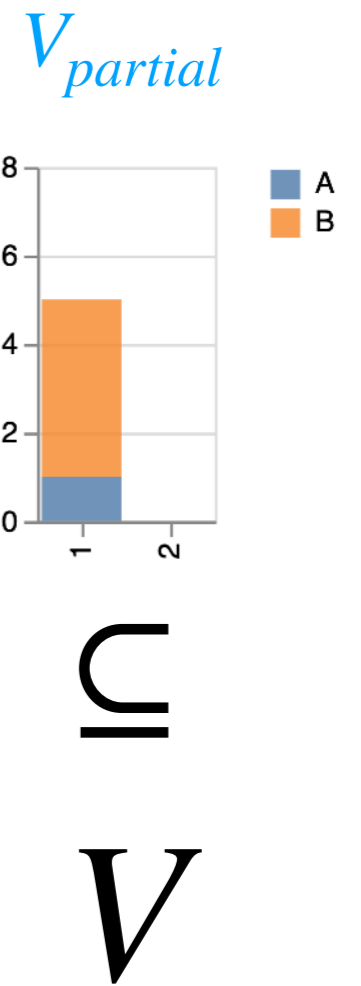
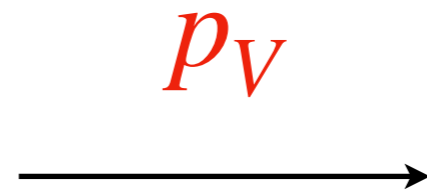
# Task decomposition

$T$

Group	A	B	C
1	0	2	1
1	1	2	3
2	2	2	0
2	1	4	1



$T'$





# Task decomposition

**Task 2: Synthesize data transformation**

Objective:  $p_T(T) \supseteq T^*$

$T$

Group	A	B	C
1	0	2	1
1	1	2	3
2	2	2	0
2	1	4	1

$p_T$

**Task 1: Decompile visualization**

$p_V(T^*) = V_{partial}$

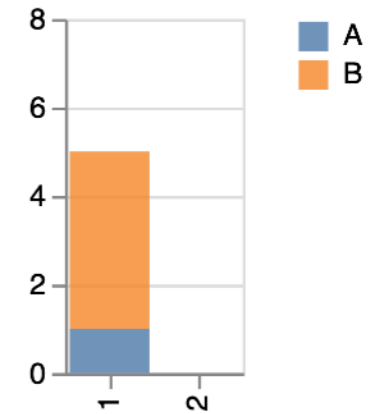
$T^*$

C1	C2	C3
1	A	1
1	B	4

$p_V$

`bar(x ← C1,  
height ← C3,  
color ← C2)`

$V_{partial}$



$\subset$

$T'$

$\cup$

$V$

## Step 2: Synthesize Data Transformation

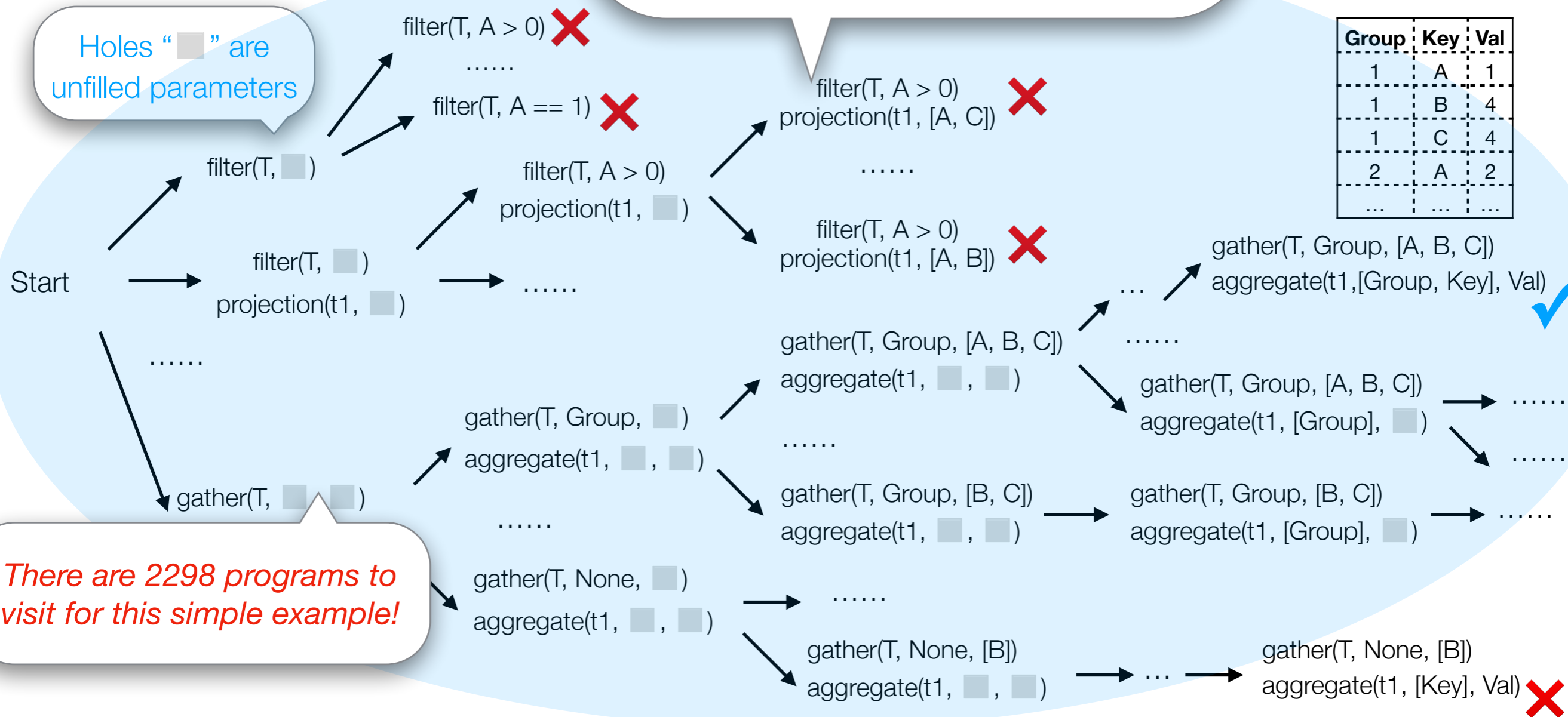
$T$				$T^*$		
Group	A	B	C	C1	C2	C3
1	0	2	1	1	A	1
1	1	2	3	1	B	4
2	2	2	0			
2	1	4	1			

Objective: synthesize  $p_T$  such that  $T^* \subseteq p_T(T)$

How can we make this combinatorial search process tractable?

gather, spread, separate, ...  
 filter, join, filter, select, ...  
 aggregate, .....

Holes "■" are unfilled parameters



Group	Key	Val
1	A	1
1	B	4
1	C	4
2	A	2
...	...	...

There are 2298 programs to visit for this simple example!

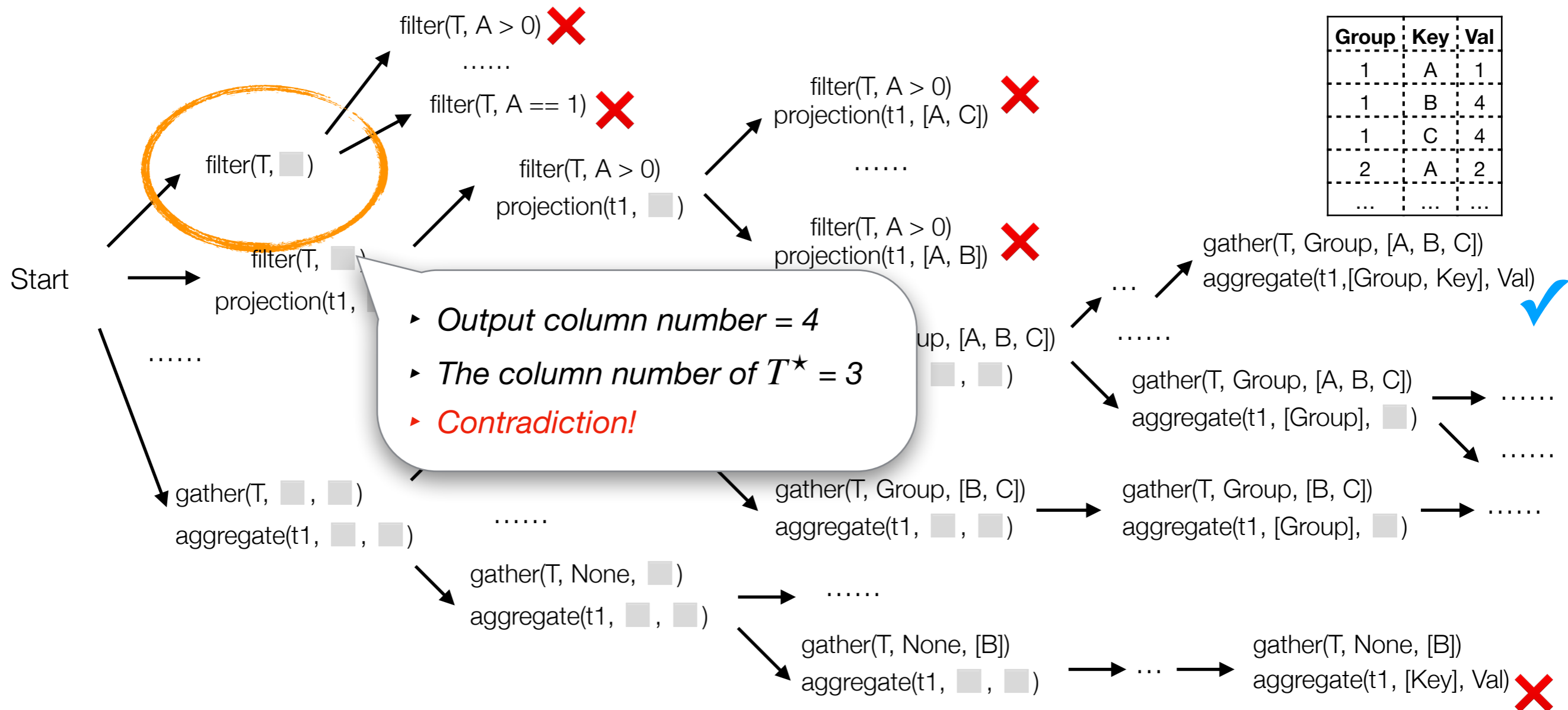
## Step 2: Synthesize Data Transformation

$T$				$T^*$		
Group	A	B	C	C1	C2	C3
1	0	2	1	1	A	1
1	1	2	3	1	B	4
2	2	2	0			
2	1	4	1			

Objective: synthesize  $p_T$  such that  $T^* \subseteq p_T(T)$

**Option: Deductive reasoning**

*Morpheus [PLDI'17, PLDI'18], Lens [POPL'20]*



## Step 2: Synthesize Data Transformation

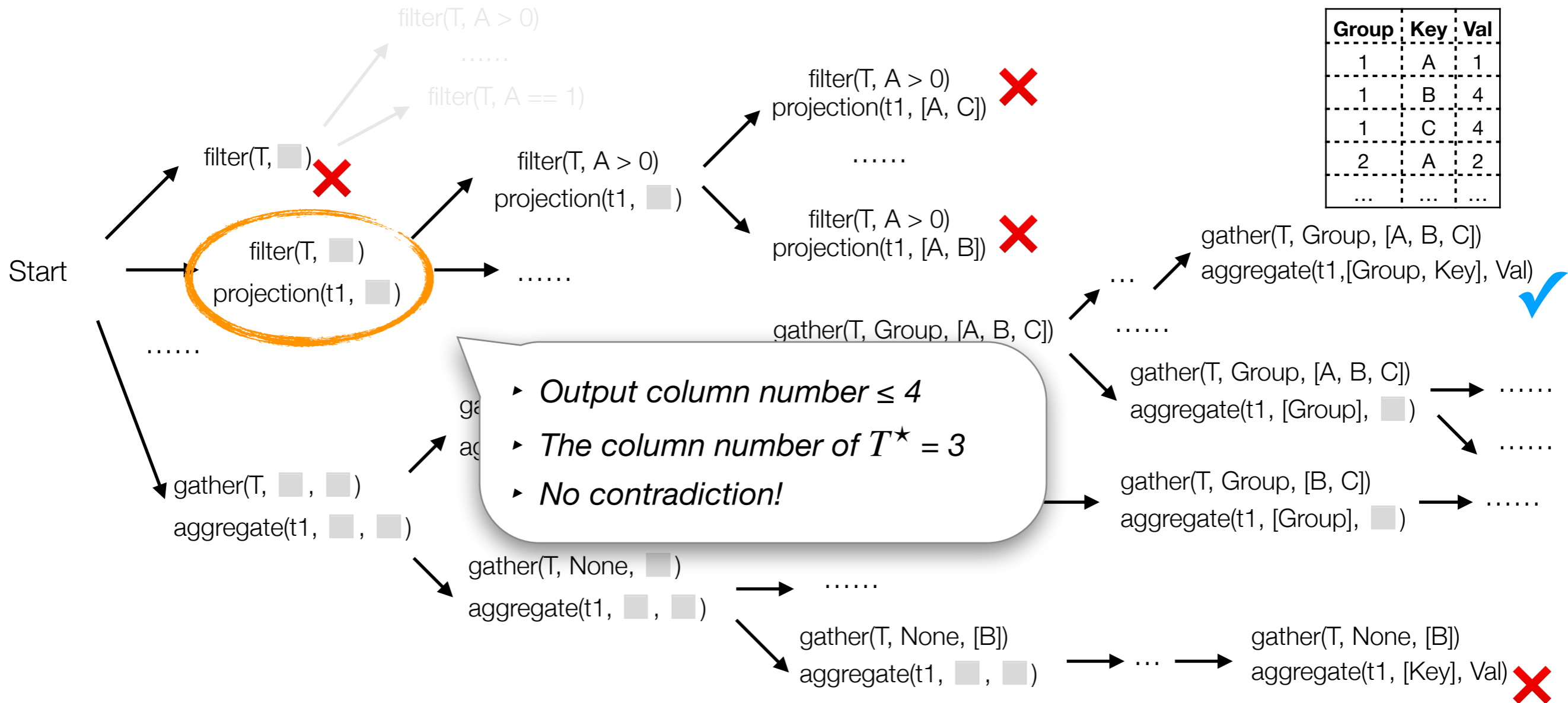
Group	A	B	C
1	0	2	1
1	1	2	3
2	2	2	0
2	1	4	1

C1	C2	C3
1	A	1
1	B	4

**Objective:  $T^* \subseteq p_T(T)$**

**Option: Deductive reasoning**  
*Morpheus [PLDI'17, PLDI'18], Lens [POPL'20]*



## Step 2: Synthesize Data Transformation

Group	A	B	C
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C1	C2	C3
1	A	1
1	B	4

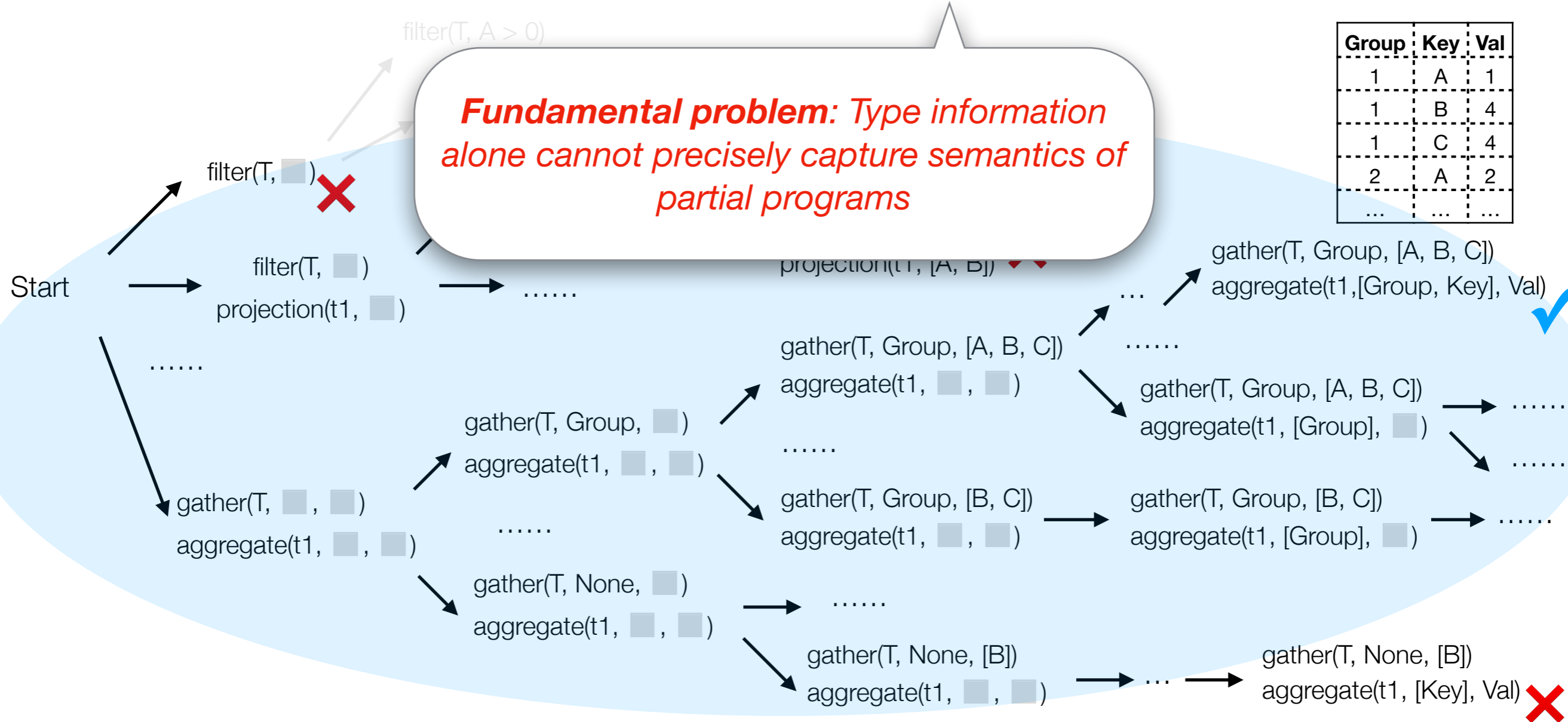
Objective:  $T^* \subseteq p_T(T)$

Option: Deductive reasoning

Morpheus [PLDI'17, PLDI'18], Lens [POPL'20]

Fundamental problem: Type information alone cannot precisely capture semantics of partial programs

Group	Key	Val
1	A	1
1	B	4
1	C	4
2	A	2
...	...	...



## Step 2: Synthesize Data Transformation

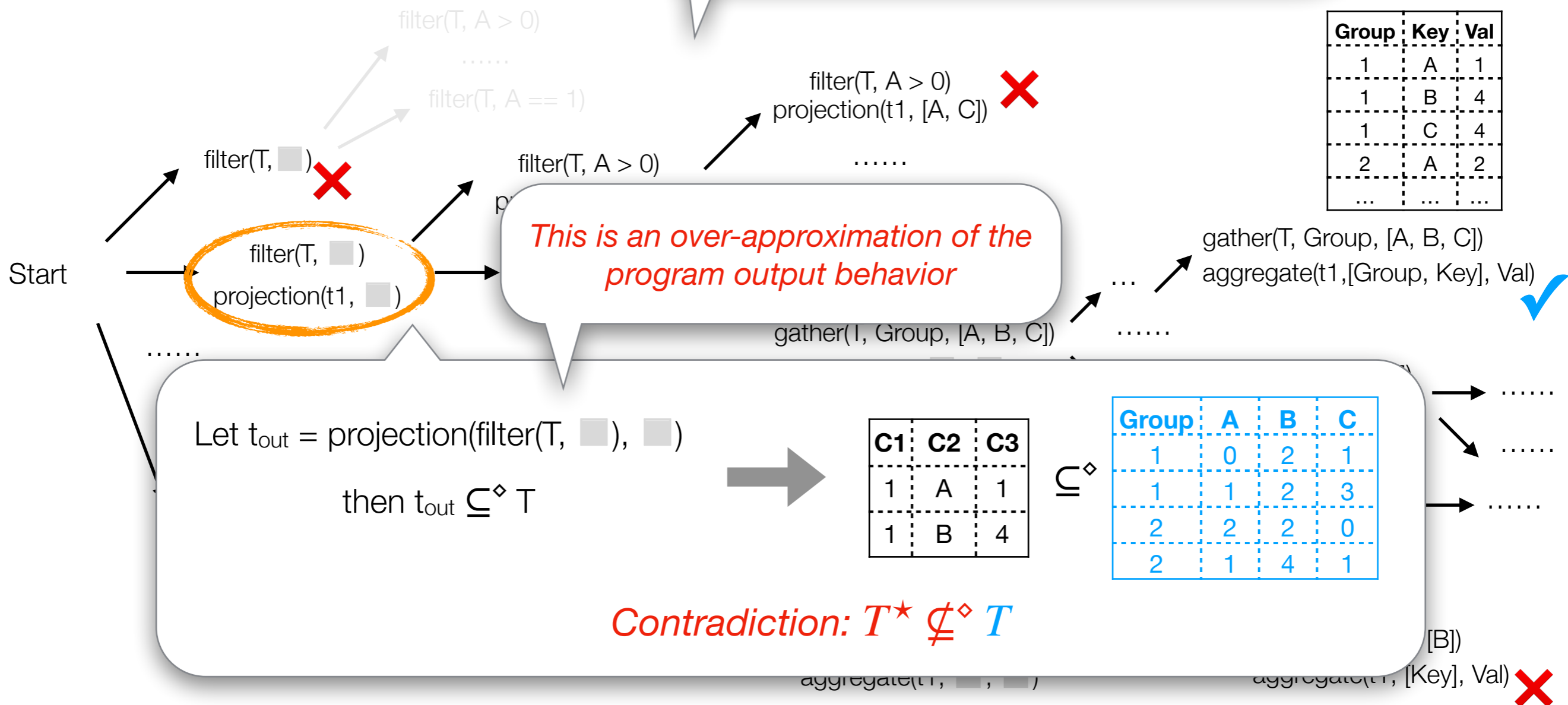
Group	A	B	C
1	0	2	1
1	1	2	3
2	2	2	0
2	1	4	1

C1	C2	C3
1	A	1
1	B	4

Objective:  $T^* \subseteq p_T(T)$

**Reasoning with abstract semantics**

Propagate concrete values through partial programs to analyze their behaviors







## Step 2: Synthesize Data Transformation

Group	A	B	C
1	0	2	1
1	1	2	3
2	2	2	0
2	1	4	1

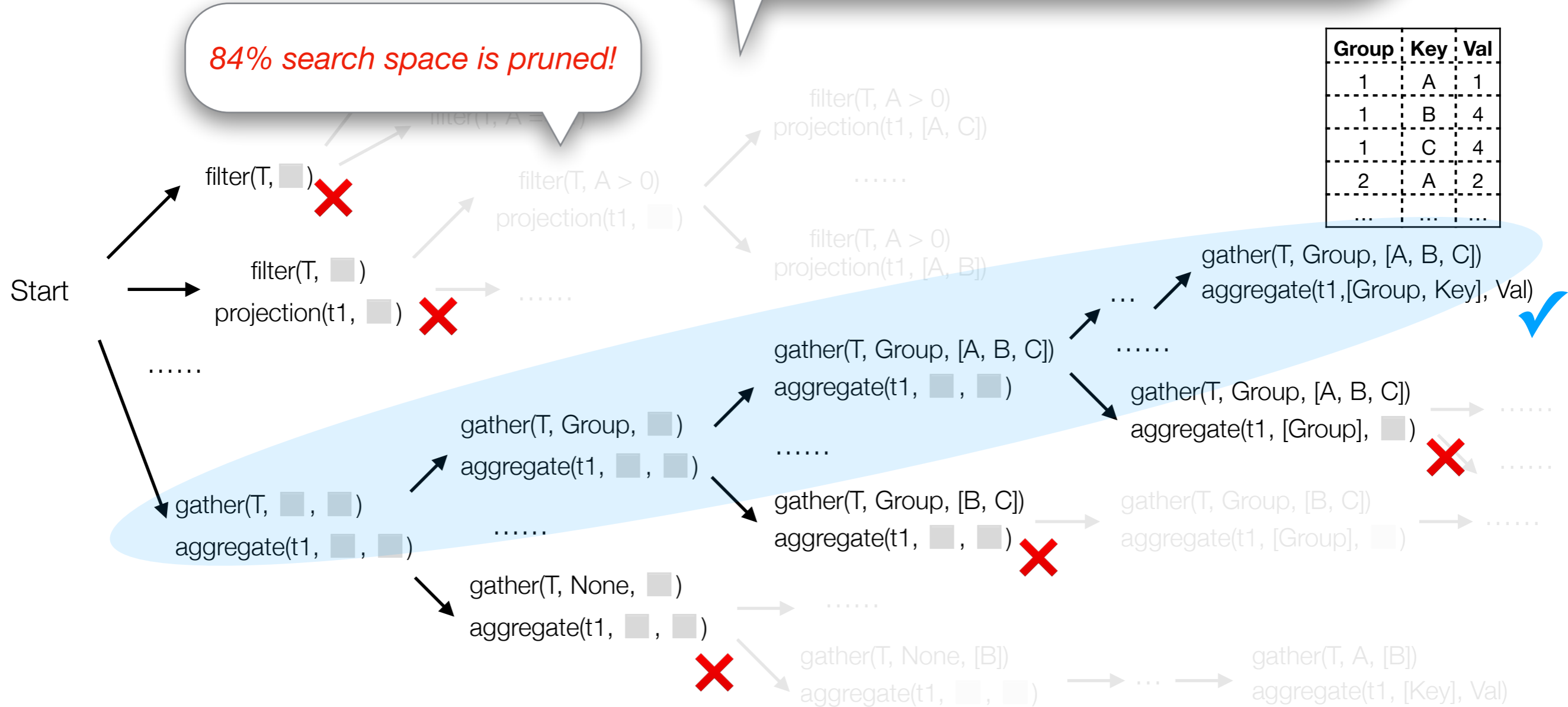
C1	C2	C3
1	A	1
1	B	4

Objective:  $T^* \subseteq p_T(T)$

**Reasoning with abstract semantics**  
 Propagate concrete values through partial programs to analyze their behaviors

[PLDI 17, POPL 20]

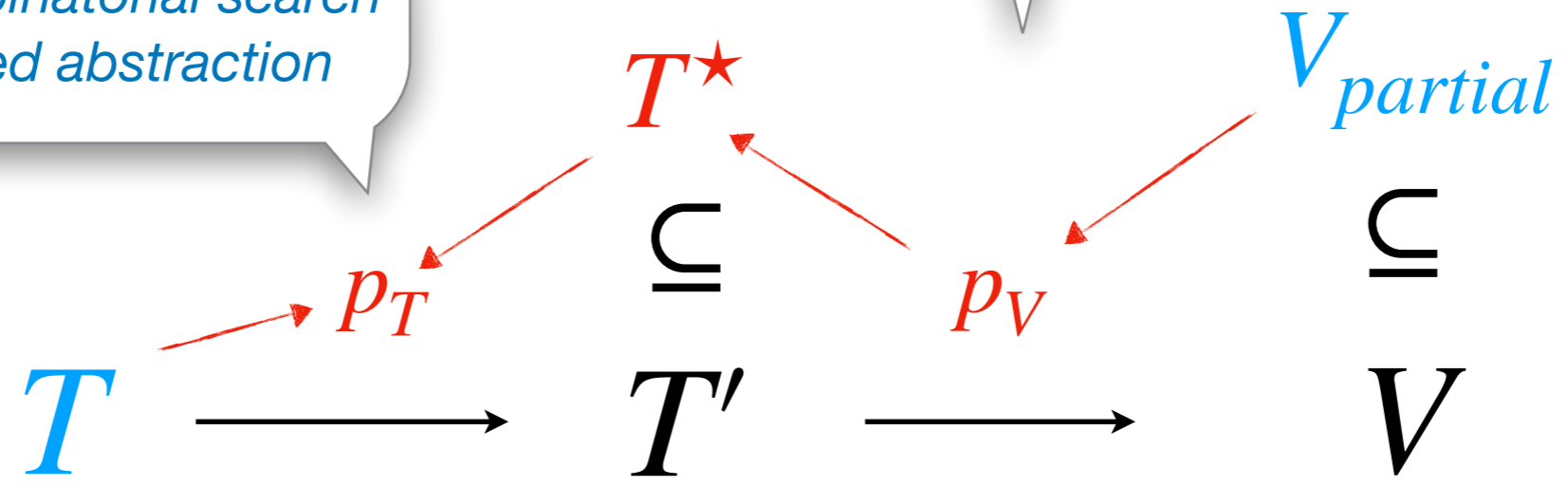
84% search space is pruned!



# Synthesis algorithm review

2. Scale up combinatorial search with value-based abstraction

1. Decompose the task with visualization decompilation



## Language support

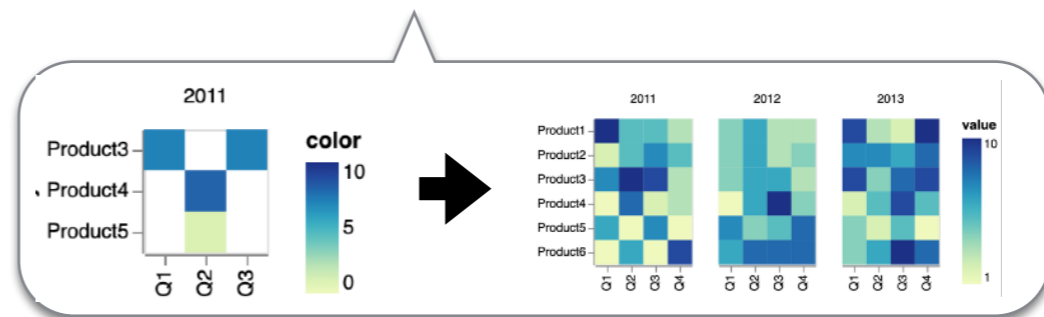
- R tidyverse
- Vega-Lite

## 83 benchmarks

- Stack Overflow
- Excel / R tutorials

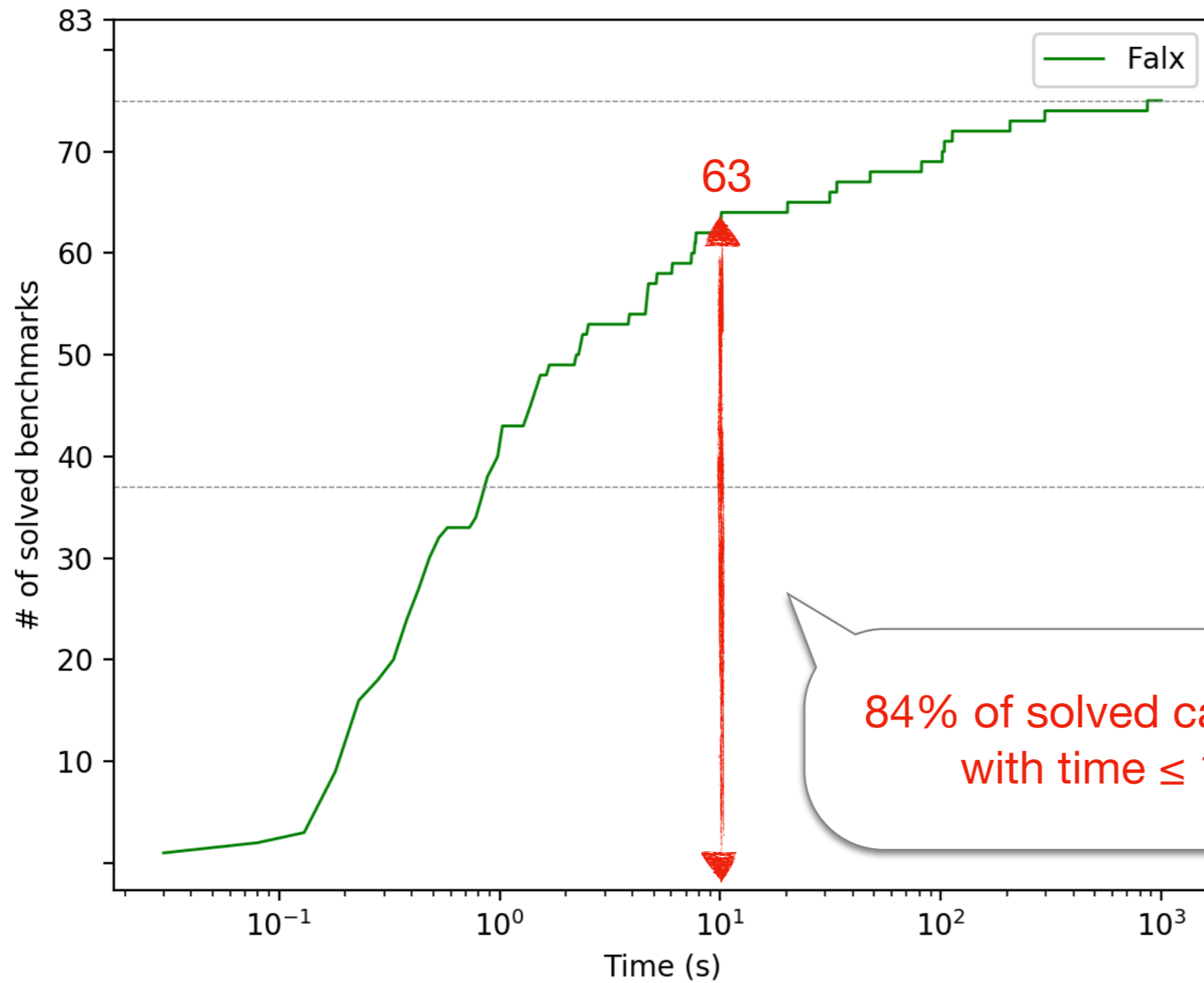
## Process

- Test synthesis time with partial visualizations sampled from full visualizations.



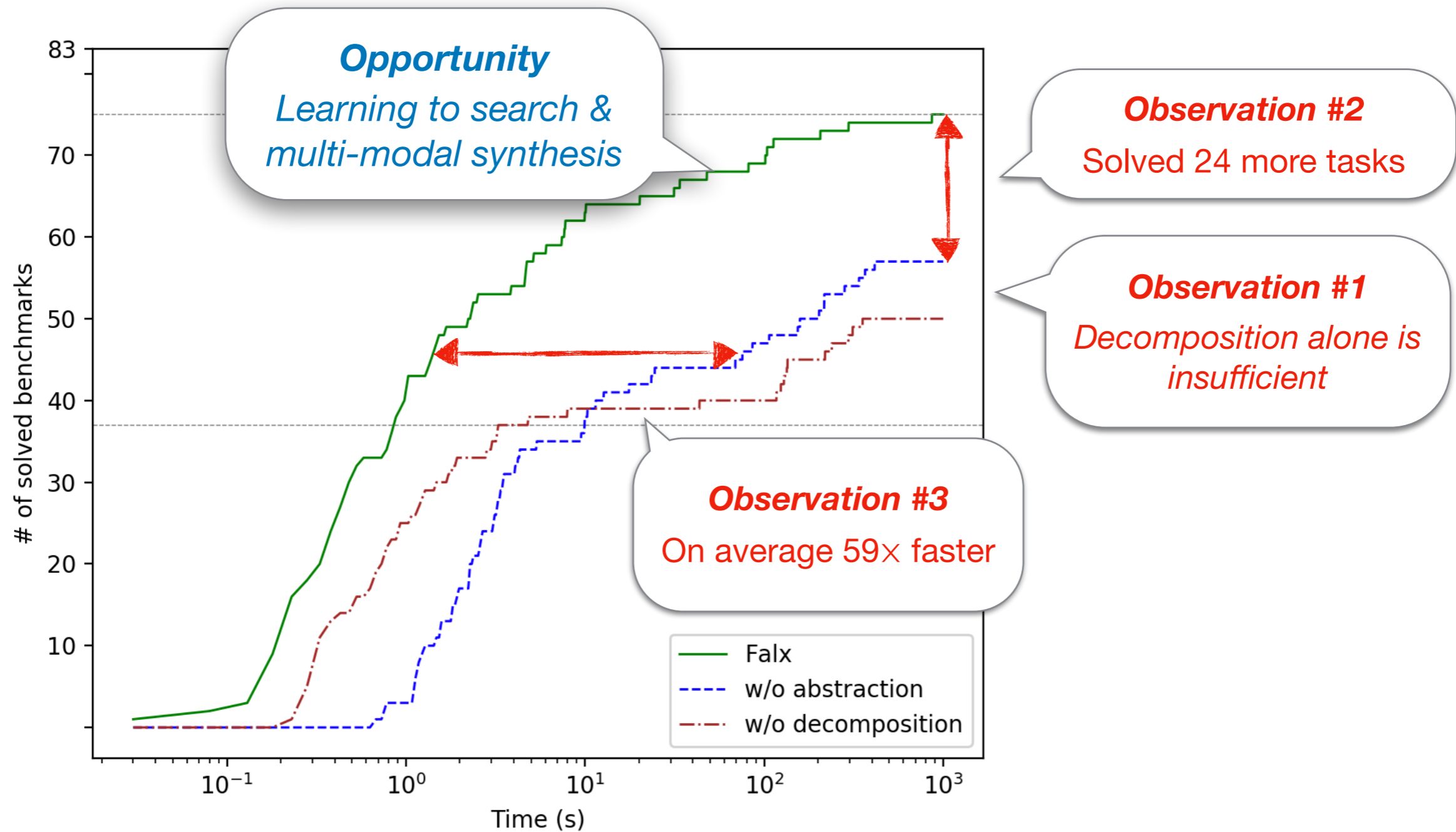
# Can Falx solve practical visualization tasks?

Evaluation with example size = 4



# Does abstraction and decomposition play important roles in Falx?

Evaluation with example size = 4

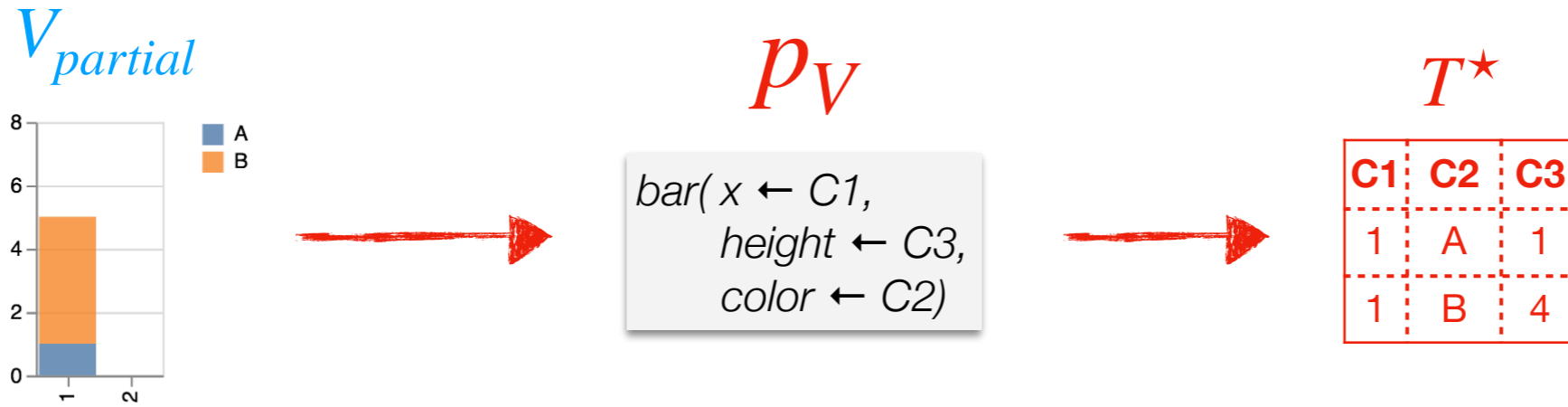


Number of benchmarks that can be solved within given time limit (log scale)

# What I didn't cover:

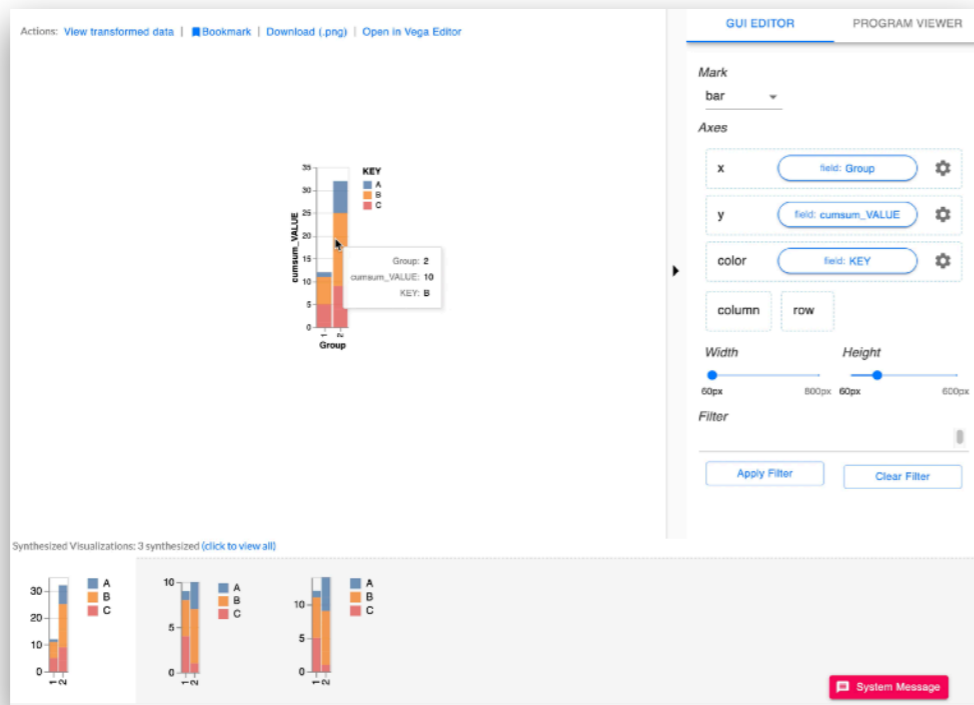
Visualization by Example [POPL 20]

(1) How to design the language  $\mathcal{L}_V$  to support easy decompilation



(2) Interactive model design

(e.g., how to create example / how to handle ambiguity)



Falx: Synthesis-powered Visualization Authoring [CHI 21 Best Paper]

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