

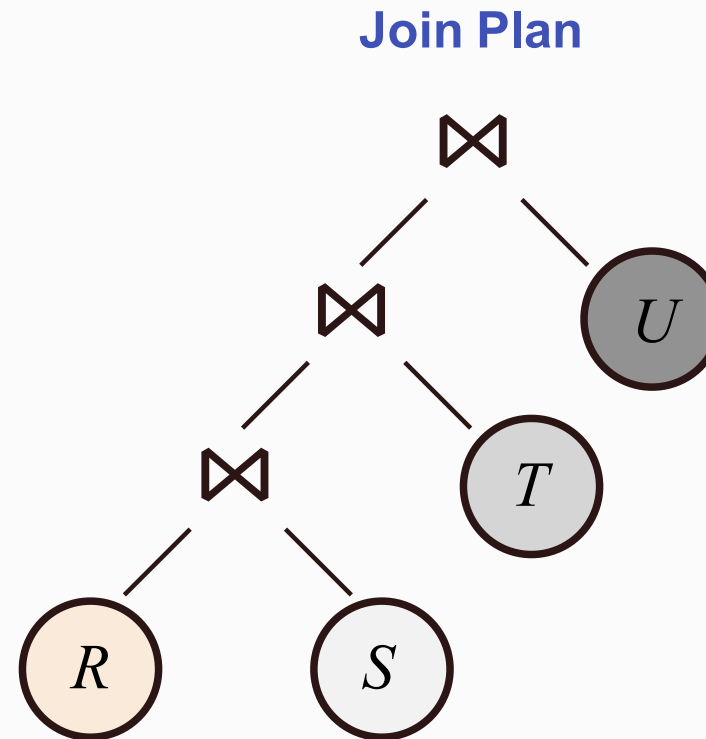
Robust Predicate Transfer With Dynamic Execution

Yiming Qiao¹, Peter Boncz², Huanchen Zhang¹

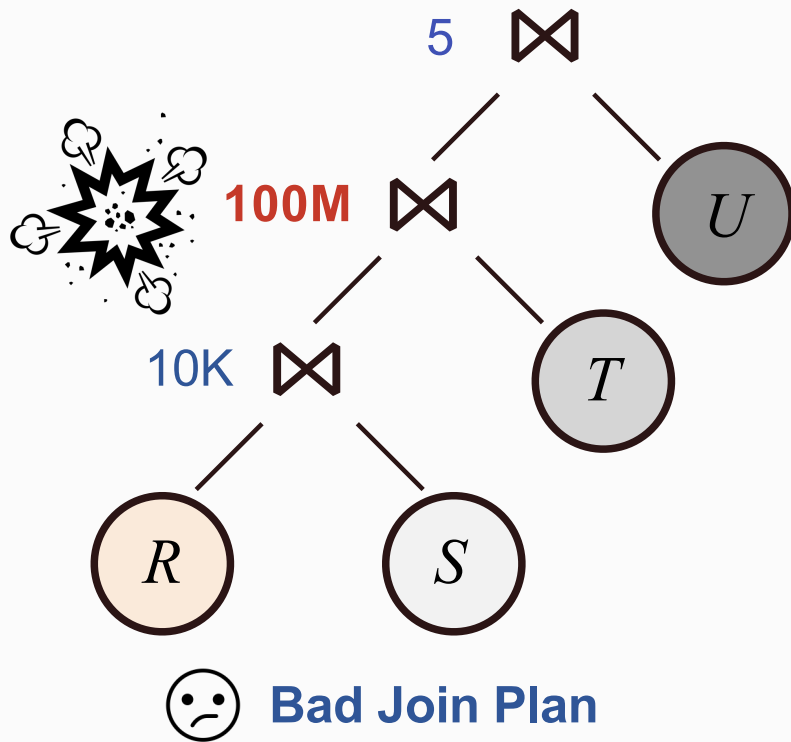
¹Tsinghua University ²CWI

Join Order Matters

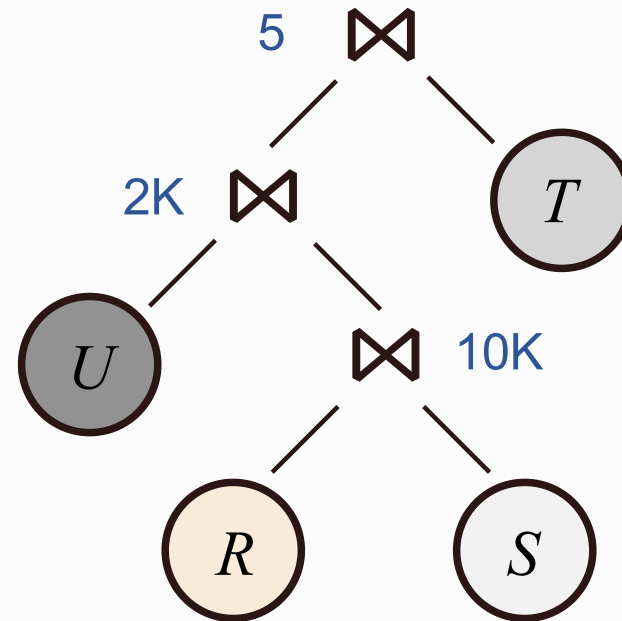
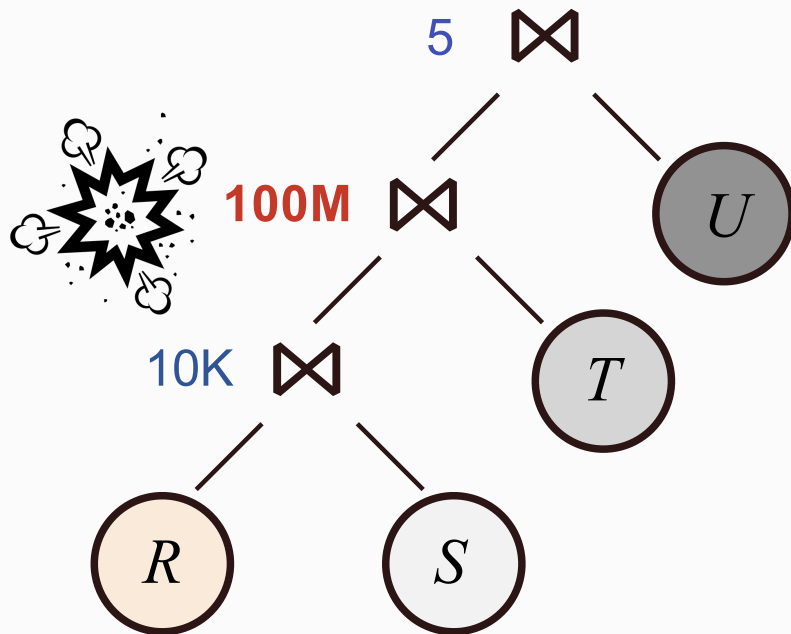
```
SELECT *  
FROM   R, S, T, U  
WHERE  R.a = S.a  
AND    R.b = T.b  
AND    R.c = U.c
```



Join Order Matters

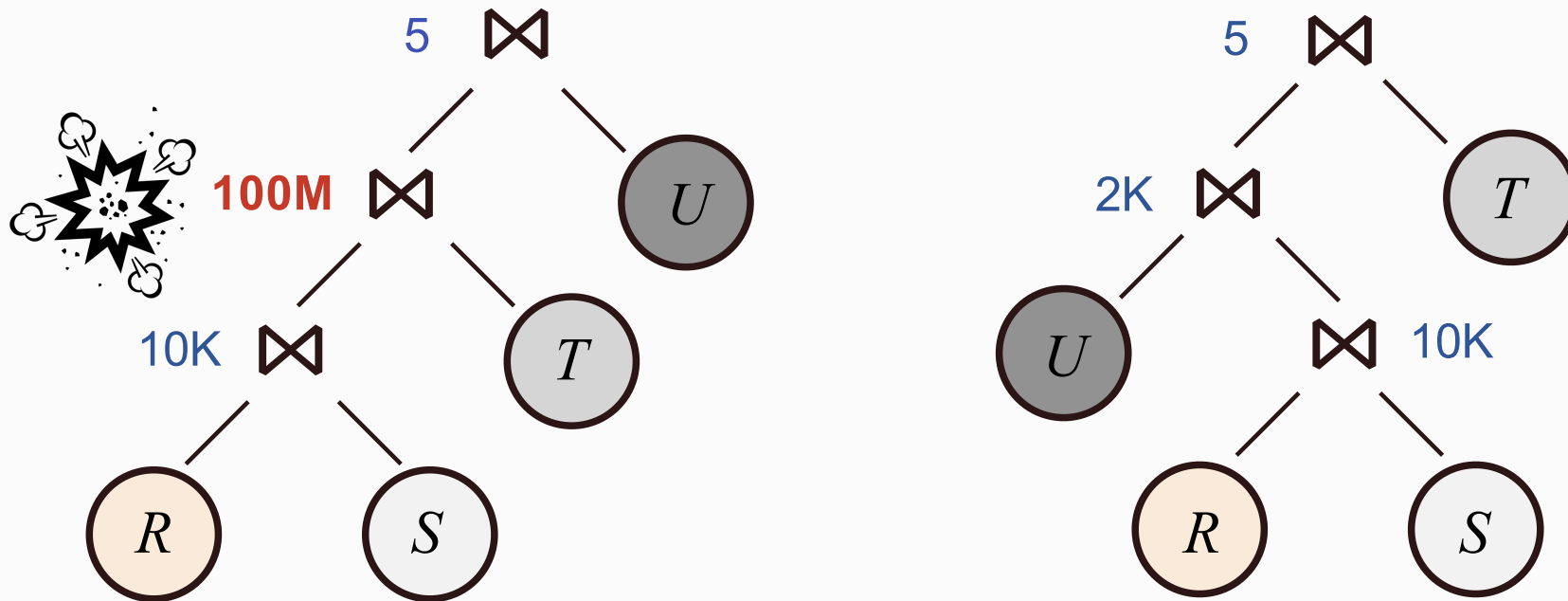


Join Order Matters



😊 Nice Join Plan

Join Order Matters



Much prior work studies how to choose the join plan with minimum intermediate result size

From One Join Equality to Two Filters

```
SELECT *  
FROM    R, S, T, U  
WHERE   R.a = S.a  
AND     R.b = T.b  
AND     R.c = U.c
```

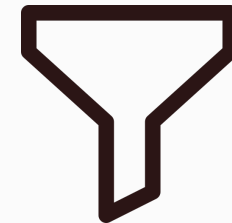
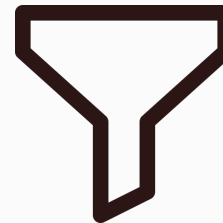
From One Join Equality to Two Filters

WHERE $R.a = S.a$ **=** $R.a \subseteq S.a$ **and** $R.a \supseteq S.a$

From One Join Equality to Two Filters

Decompose **One Join equality** into **Two Membership Filters**

WHERE $R.a = S.a$ **=** $R.a \subseteq S.a$ **and** $R.a \supseteq S.a$



Membership Filters (e.g. Bloom Filter)

From One Join Equality to Two Filters

$R . a$

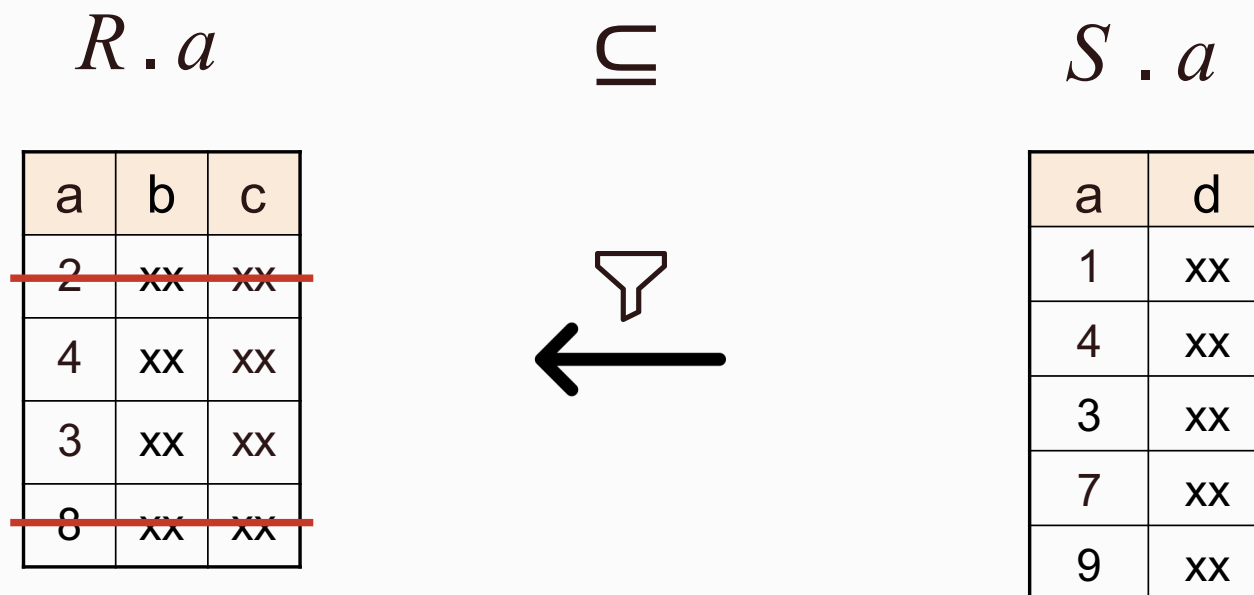
a	b	c
2	xx	xx
4	xx	xx
3	xx	xx
8	xx	xx

$\subseteq$

$S . a$

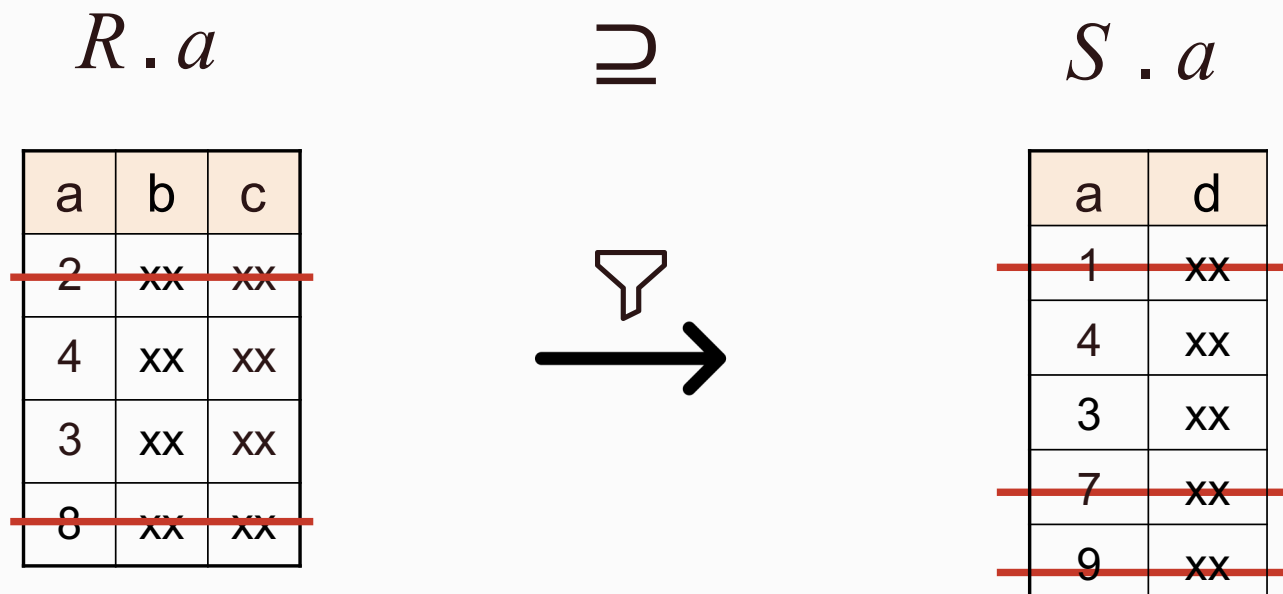
a	d
1	xx
4	xx
3	xx
7	xx
9	xx

From One Join Equality to Two Filters



Build a membership filter **based on $S.a$** to reduce the other

From One Join Equality to Two Filters



Build a membership filter **based on $R.a$** to reduce the other

From One Join Equality to Two Filters

$R . a$

a	b	c
2	xx	xx
4	xx	xx
3	xx	xx
8	xx	xx



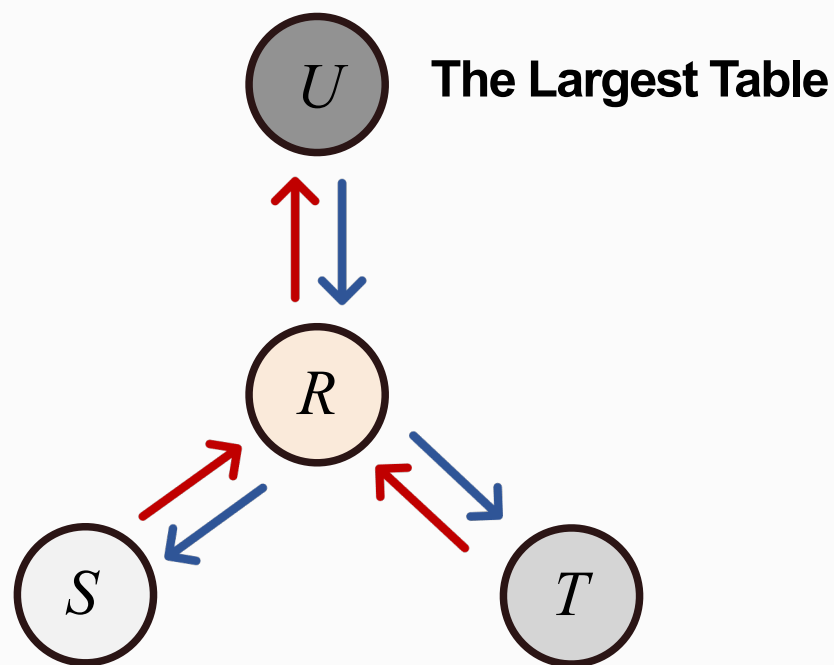
$S . a$

a	d
1	xx
4	xx
3	xx
7	xx
9	xx

Reduced Tables are then Consumed by other Relational Operators

Robust Predicate Transfer (RPT)

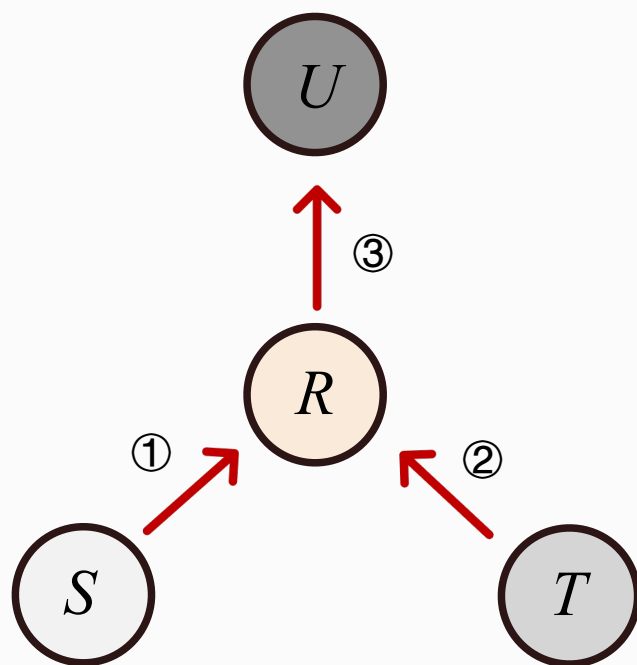
For more than two tables join, RPT introduces a transfer plan: [LargestRoot](#)



Filter Transfer Plan

Robust Predicate Transfer (RPT)

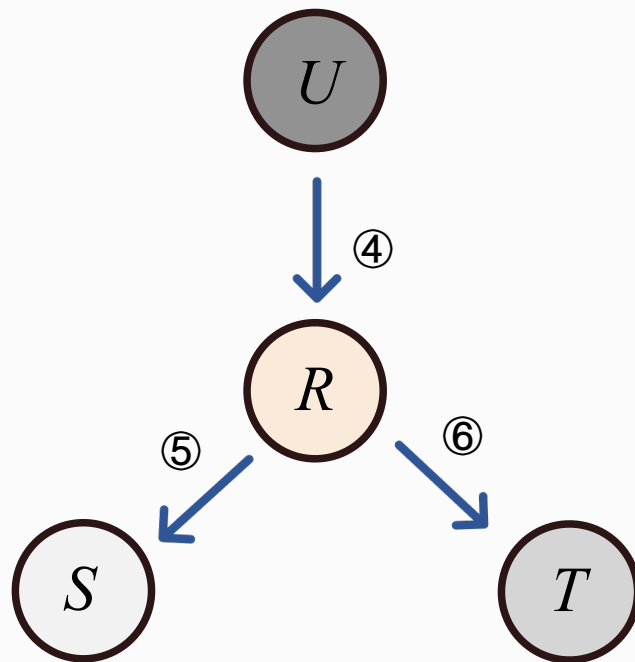
For more than two tables join, RPT introduces a transfer plan: **LargestRoot**



Forward Pass: $S \rightarrow R$, $T \rightarrow R$, $R \rightarrow U$

Robust Predicate Transfer (RPT)

For more than two tables join, RPT introduces a transfer plan: **LargestRoot**

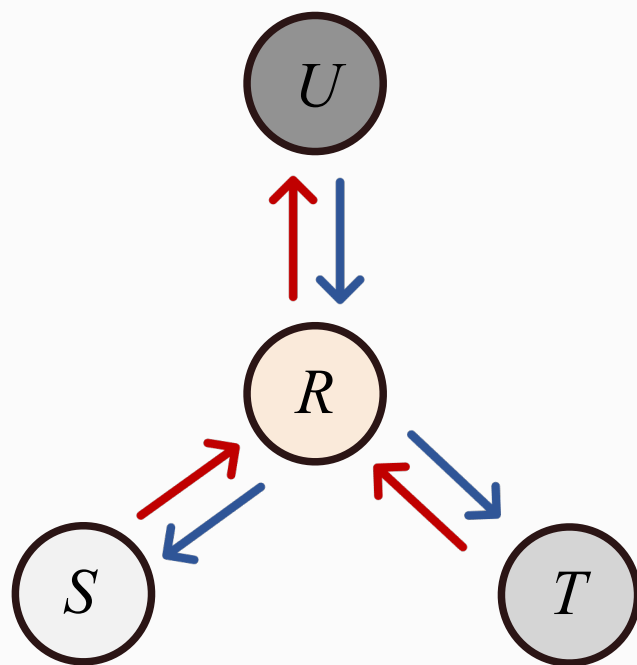


Forward Pass: $S \rightarrow R$, $T \rightarrow R$, $R \rightarrow U$

Backward Pass: $U \rightarrow R$, $R \rightarrow T$, $R \rightarrow S$

Robust Predicate Transfer (RPT)

For more than two tables join, RPT introduces a transfer plan: **LargestRoot**



Filter Transfer Plan

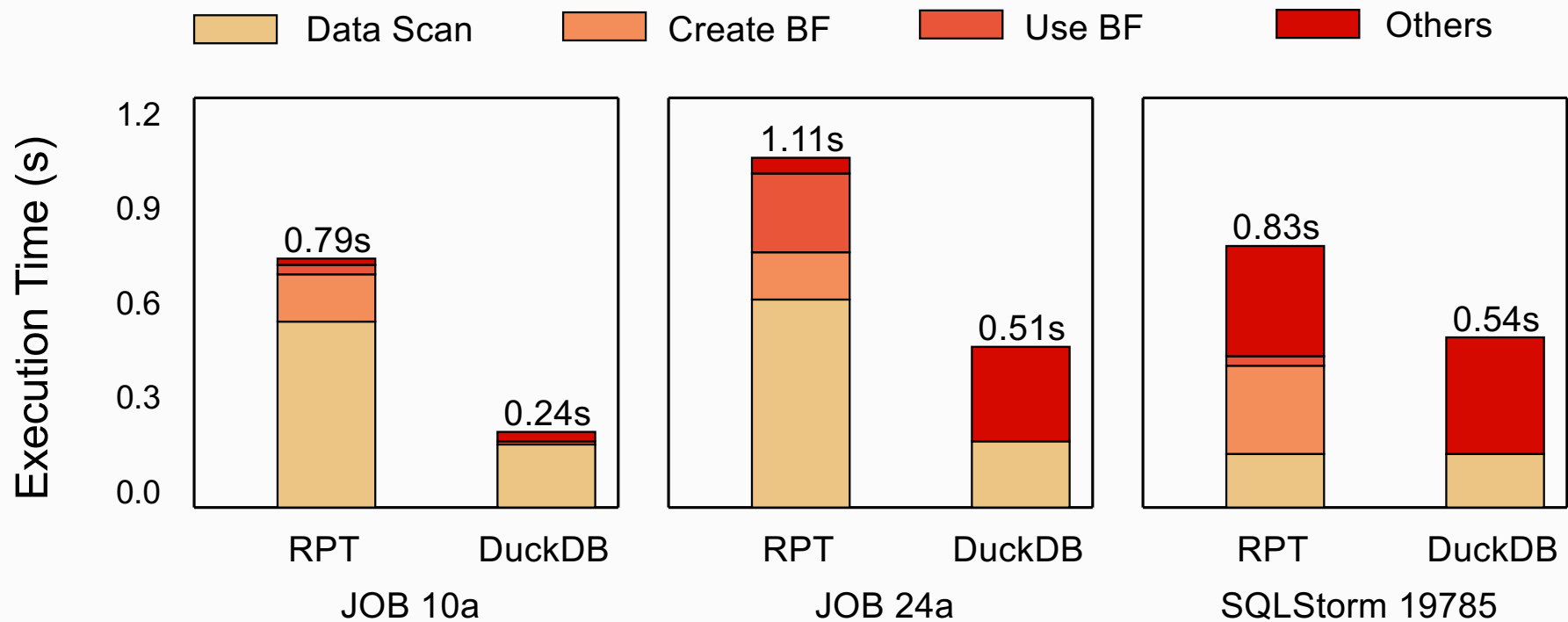
Forward Pass is from Leaves to the Root

Backward Pass is from the Root to Leaves

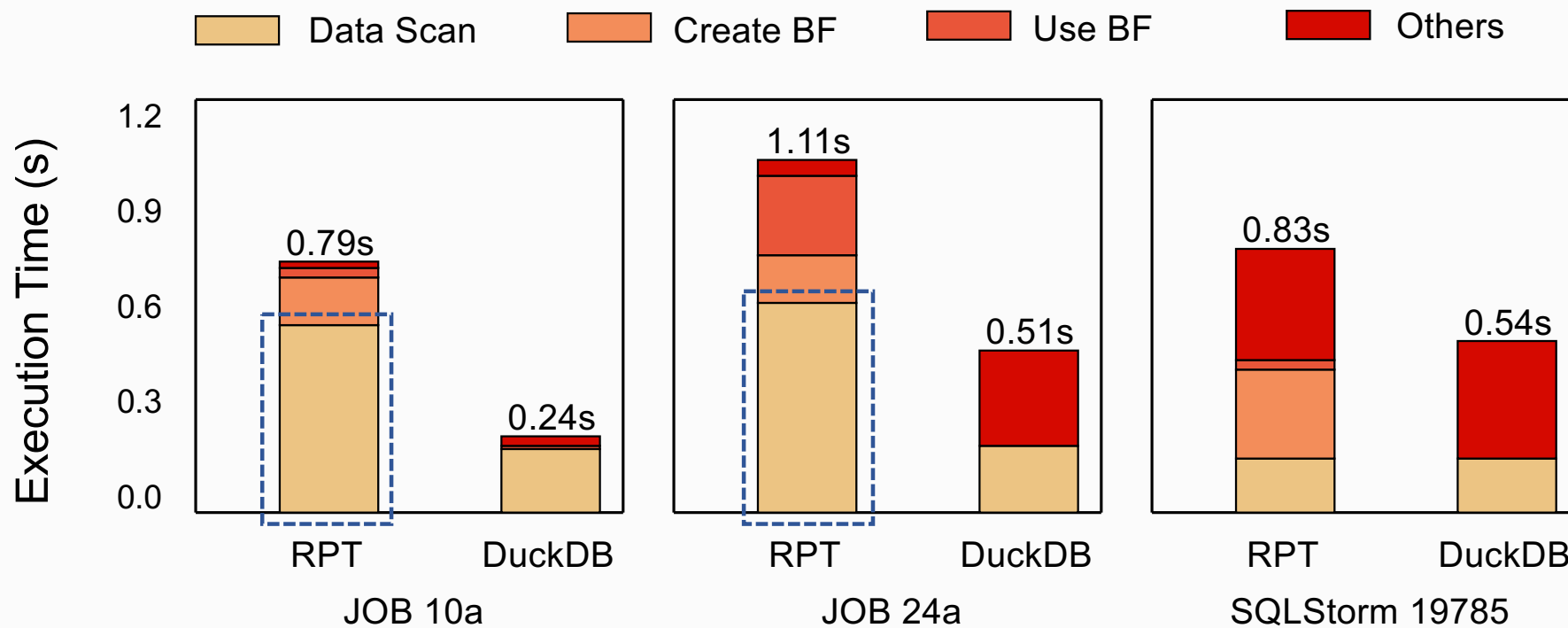
RPT preserves the robust guarantee $O(|IN| + |OUT|)$, which is **instance-optimal**, but pays

A Fixed Tax of Table Scan and Filter Transfer

Overhead of RPT

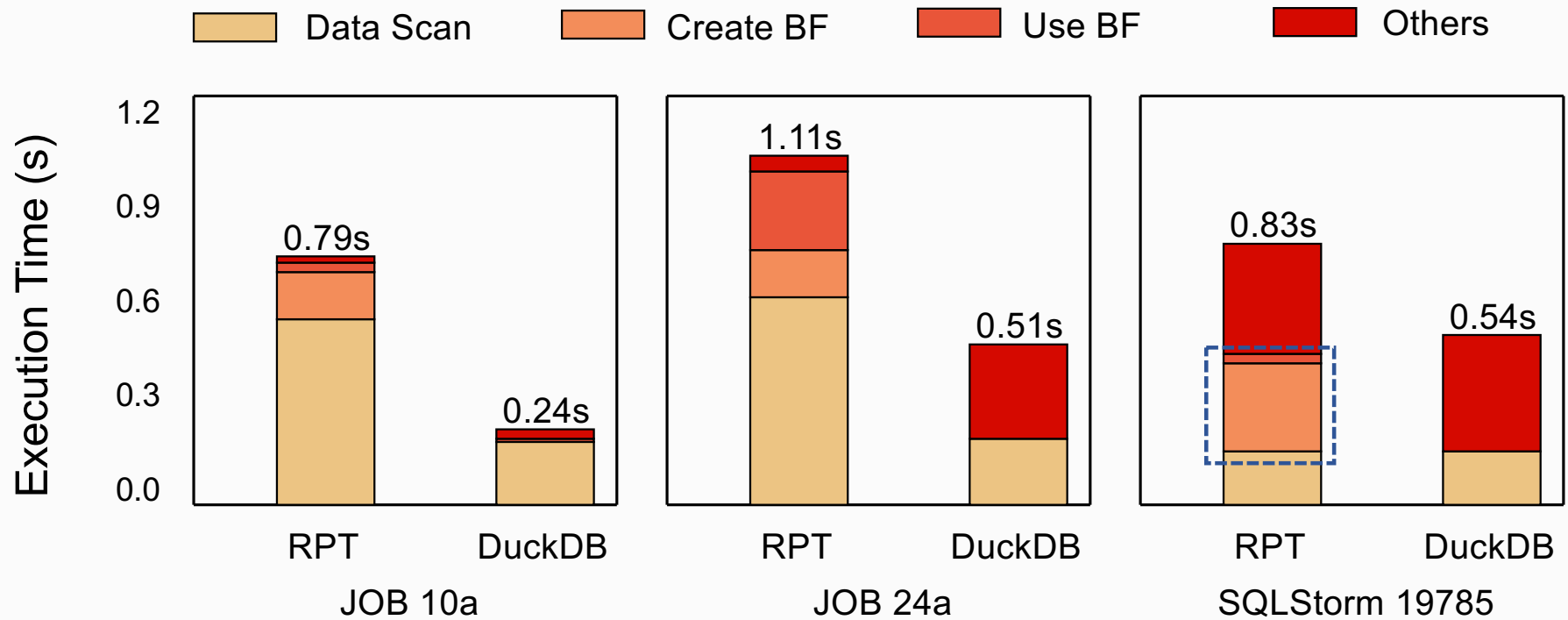


Overhead of RPT



Scan Overhead

Overhead of RPT



BF-Build Overhead

Where Does RPT Pay Overhead

1. Redundant Filter Builds

Our solution: **Asymmetric Transfer Plan**

2. Storage-Format-Unaware Filtering (e.g., row group skipping)

Our solution: **Cascade Filter**

3. Overhead on Non-Reducible Queries

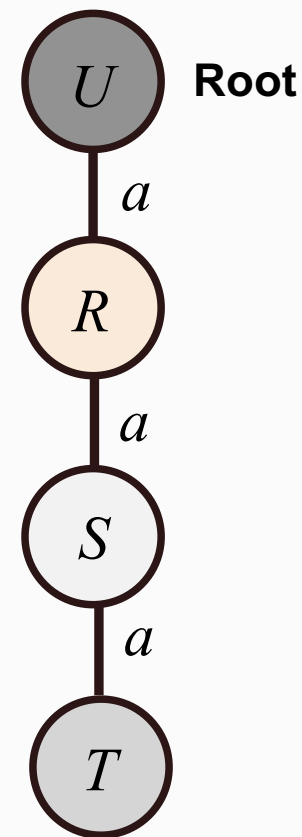
Our solution: **Dynamic Pipeline**

1. Redundant Filter Builds

```
SELECT *  
FROM   R, S, T, U  
WHERE  R.a = S.a  
AND    R.a = T.a  
AND    R.a = U.a
```

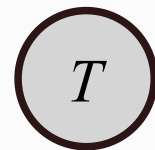
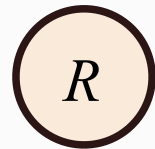
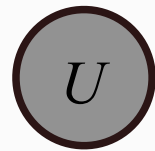
$|T| < |S| < |R| < |U|$

Largest Root (Transfer Plan)



1. Redundant Filter Builds

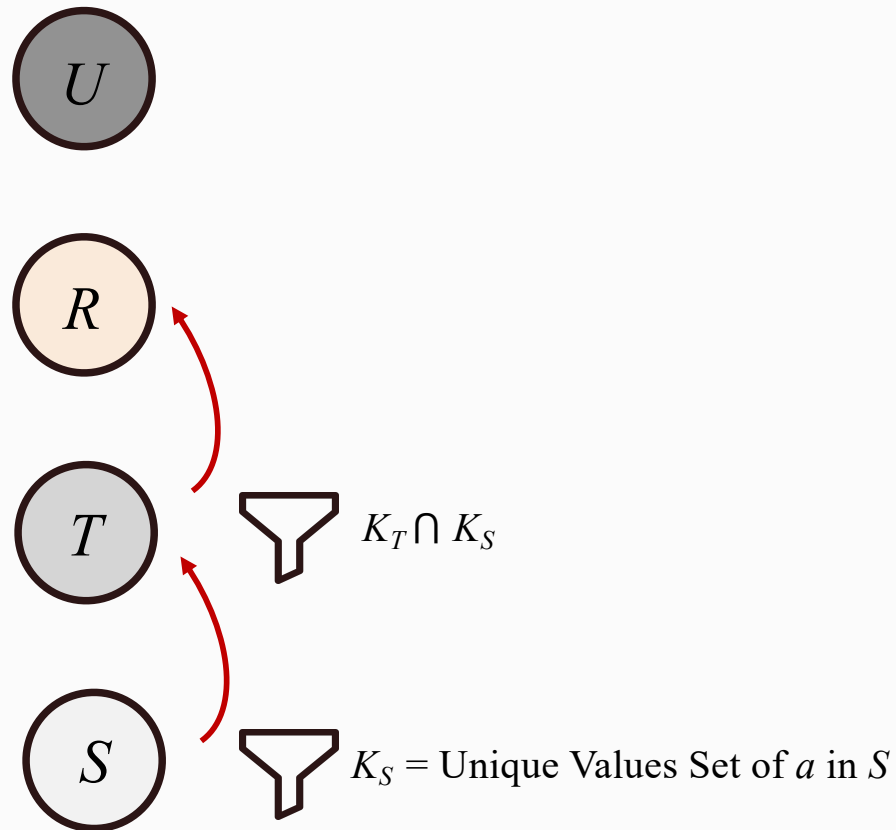
Largest Root



$K_S = \text{Unique Values Set of } a \text{ in } S$

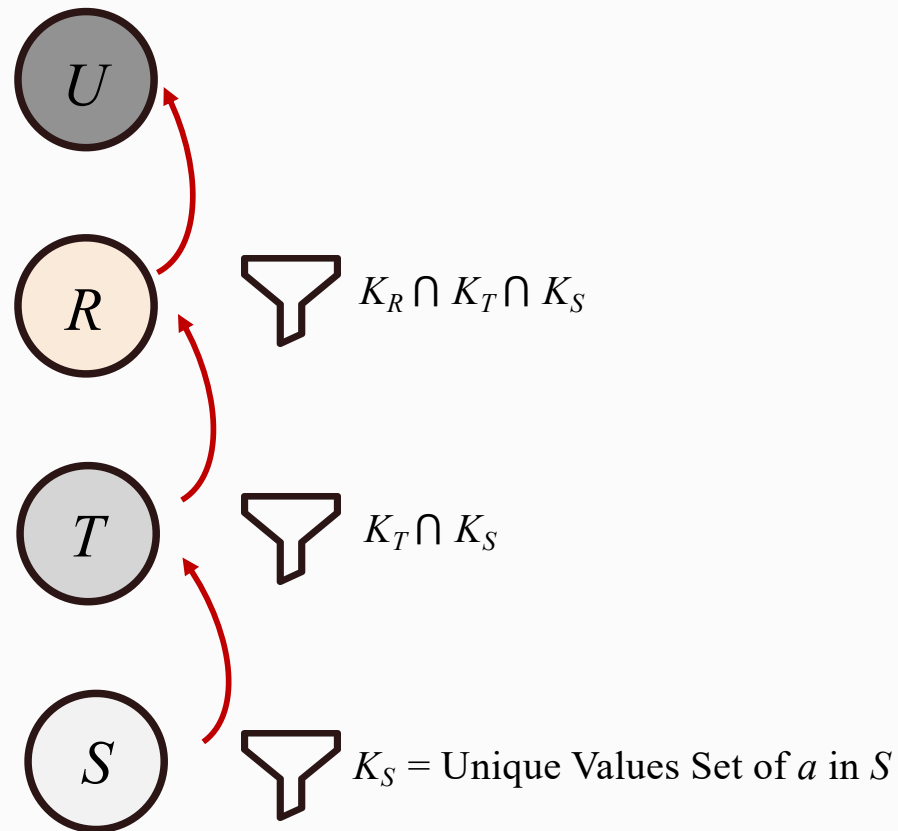
1. Redundant Filter Builds

Largest Root

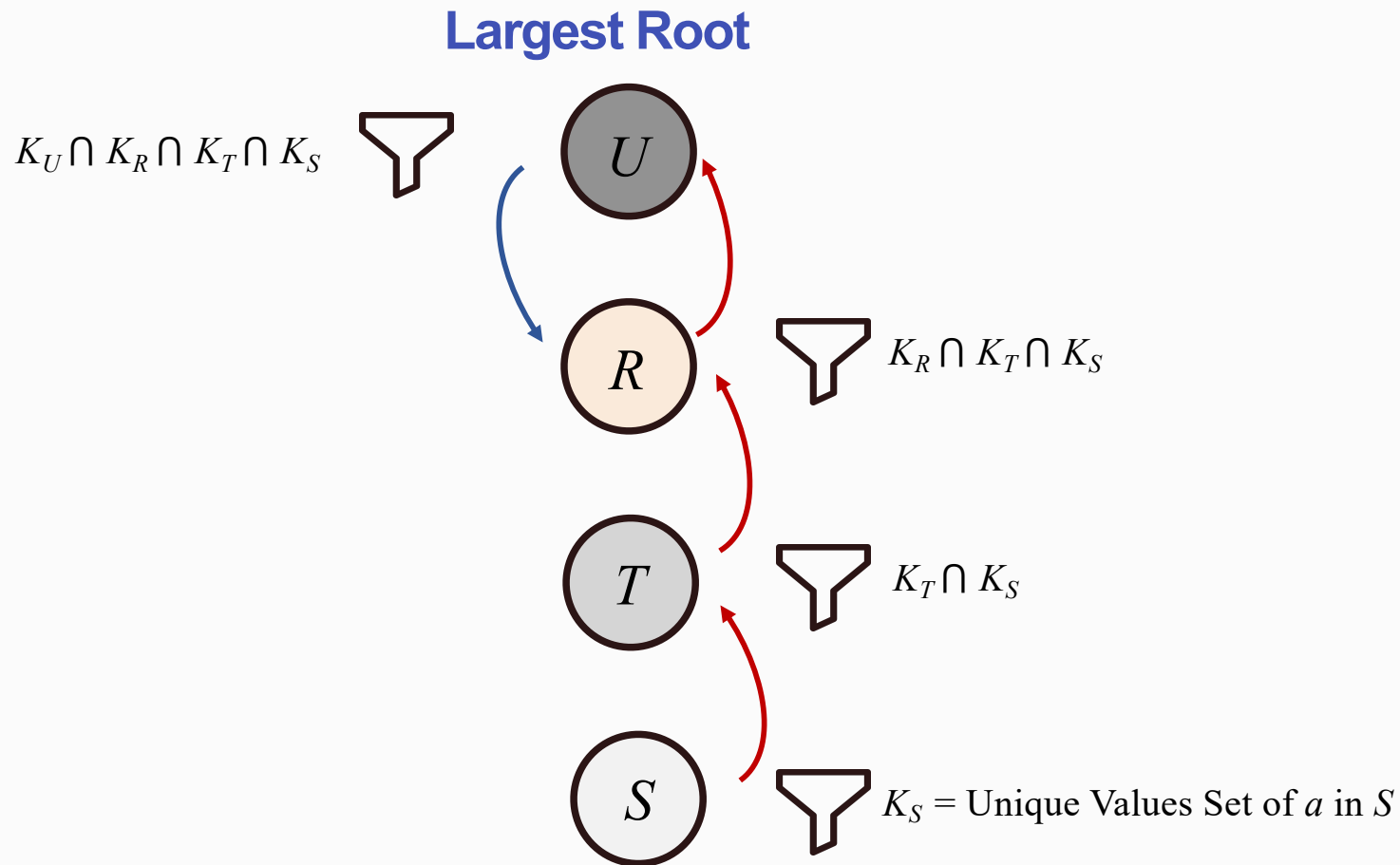


1. Redundant Filter Builds

Largest Root

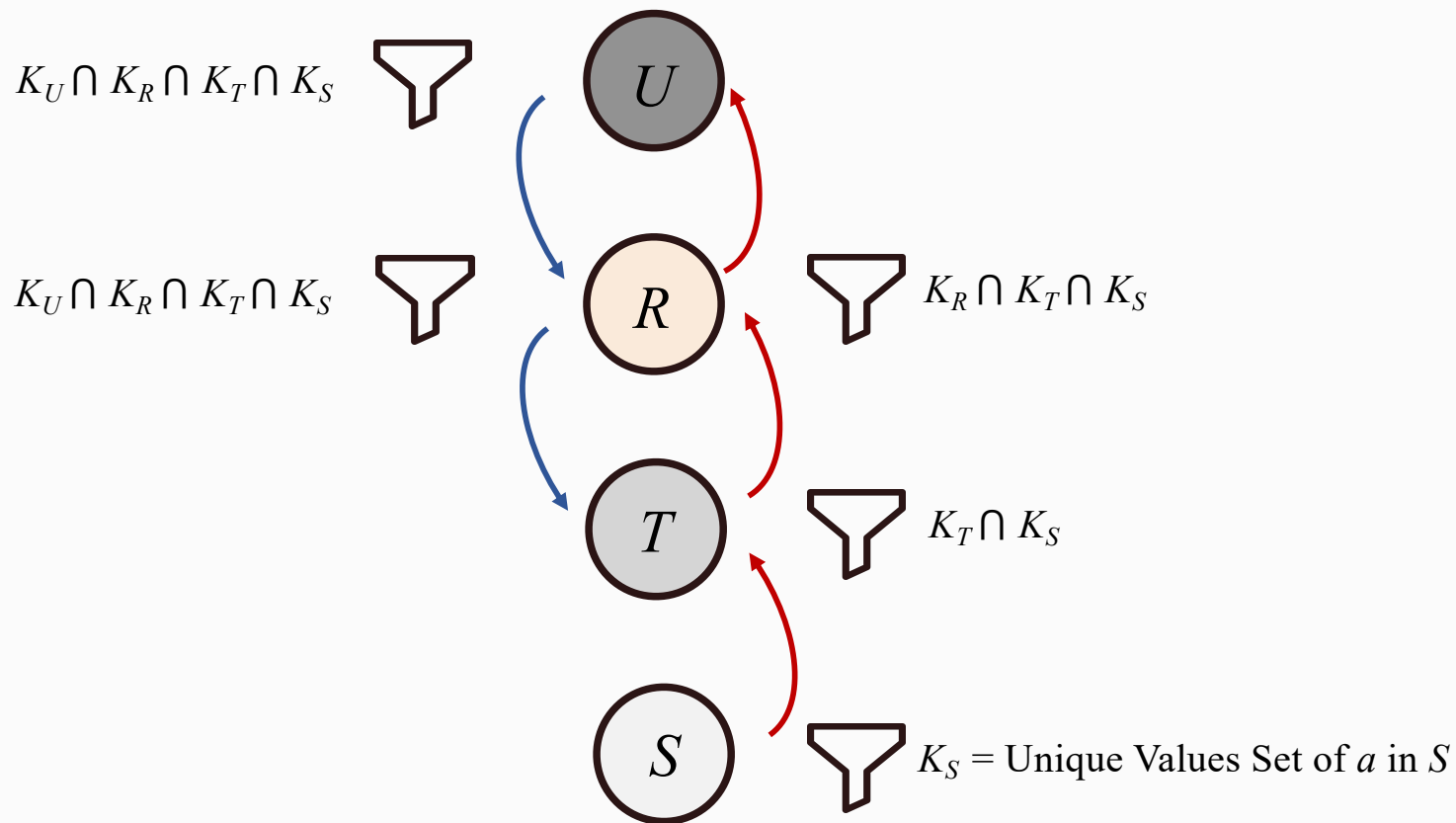


1. Redundant Filter Builds

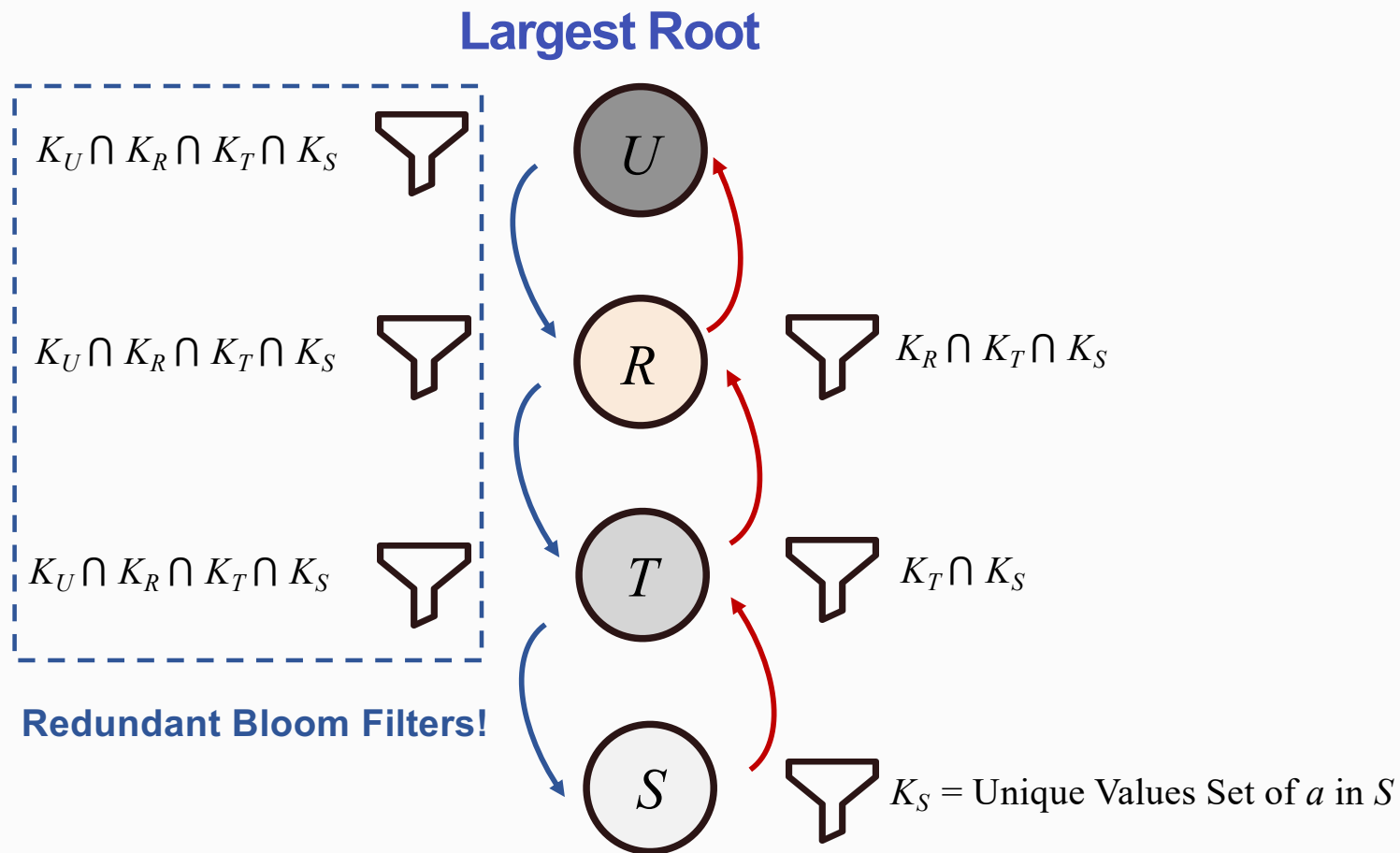


1. Redundant Filter Builds

Largest Root

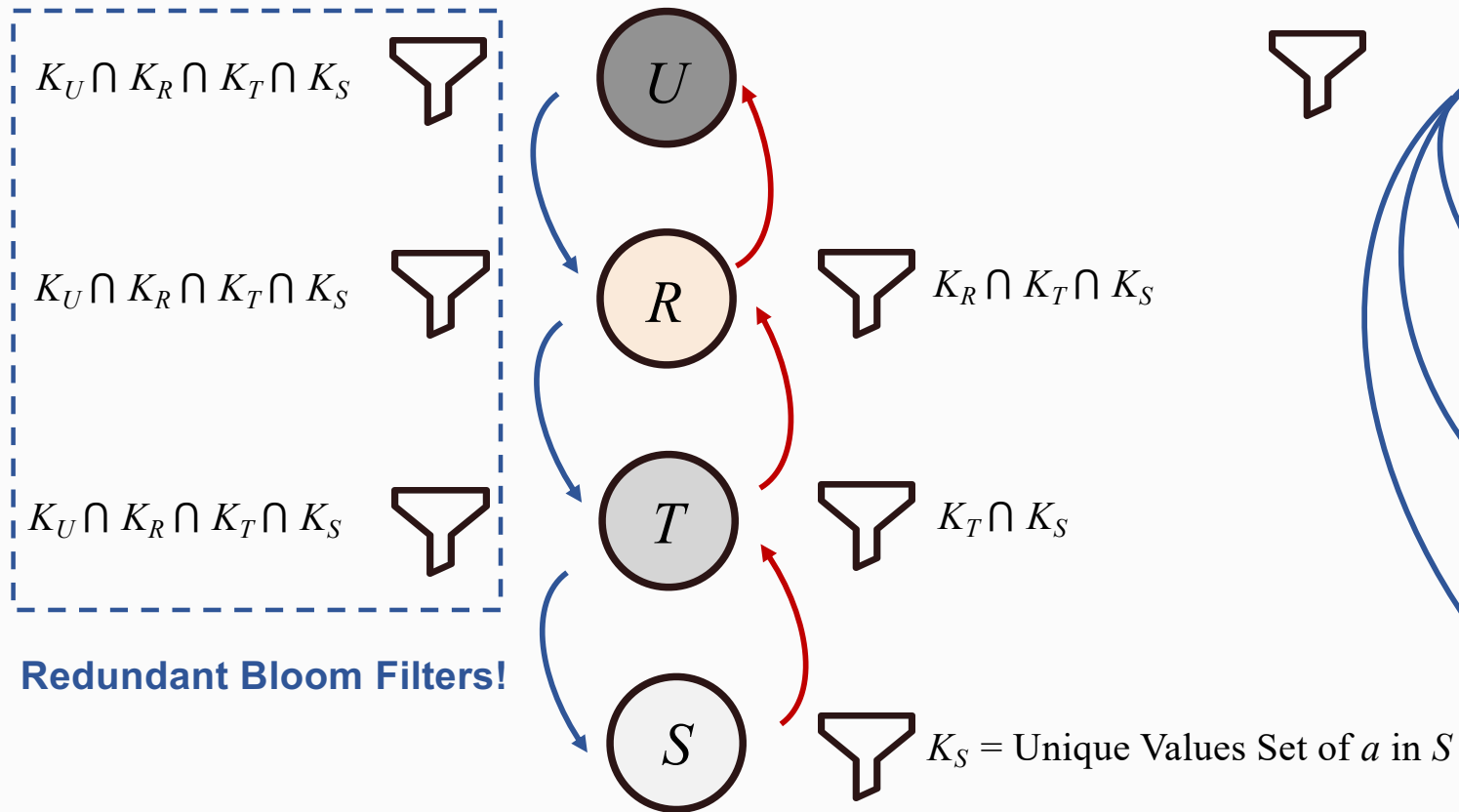


1. Redundant Filter Builds

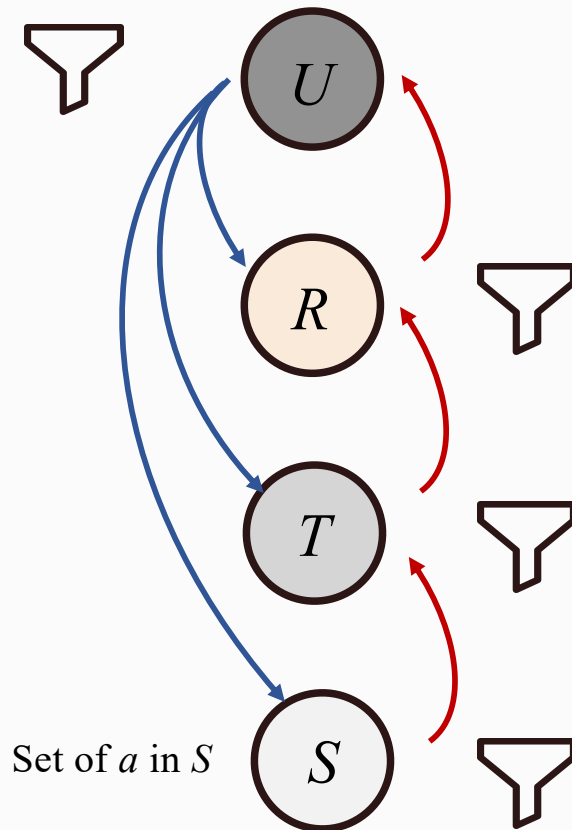


Asymmetric Transfer Plan

Largest Root

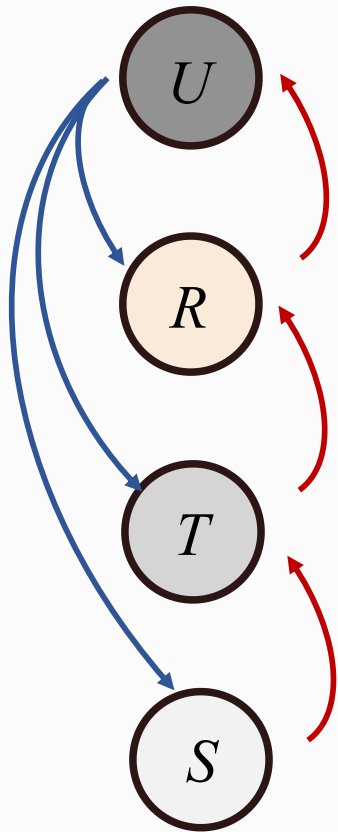


Asymmetric Transfer Plan



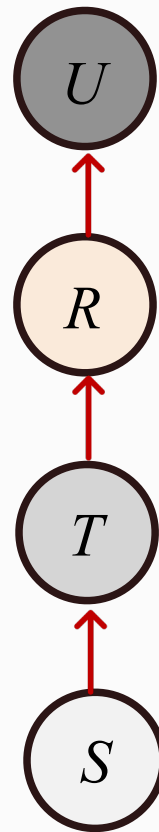
Asymmetric Transfer Plan

Asymmetric Transfer Plan



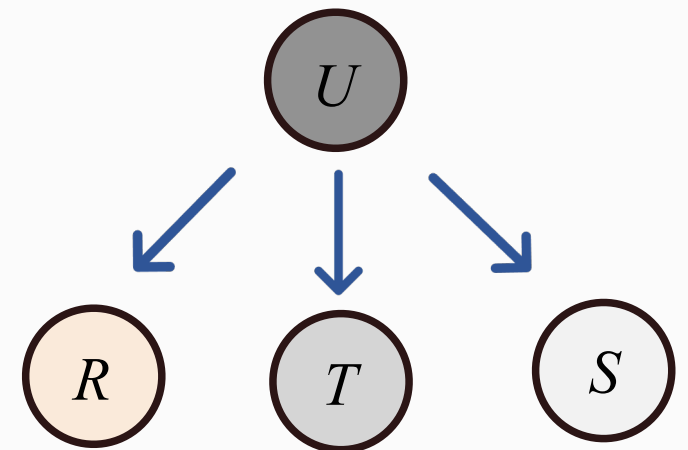
=

Forward Pass



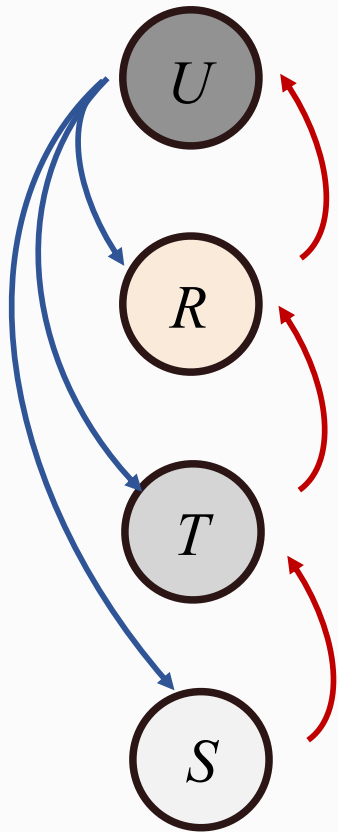
+

Backward Pass



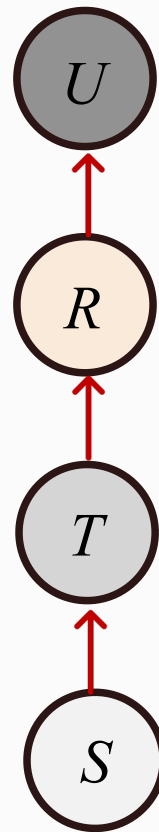
Asymmetric Transfer Plan

Asymmetric Transfer Plan



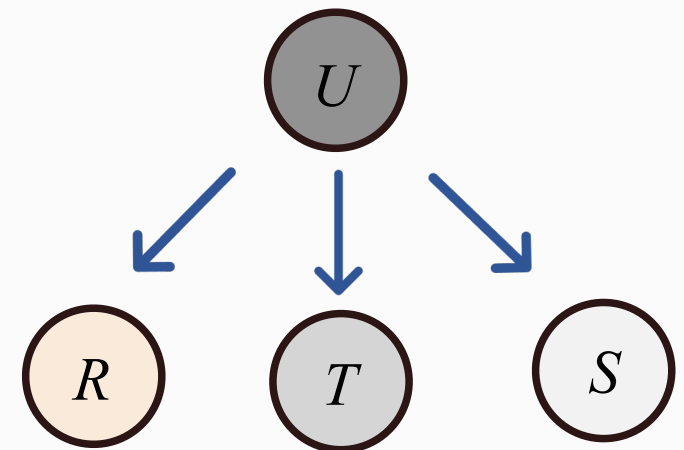
=

Forward Pass



+

Backward Pass



Yanakakis Correctness Proof:
Show they have the same root.

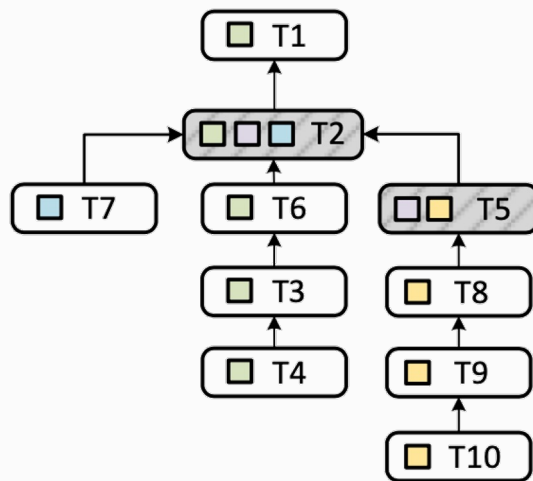
Asymmetric Transfer Plan

A deep, **chaining tree** is ideal for the forward pass,
whereas a wide, **broadcast tree** is best for the backward pass.

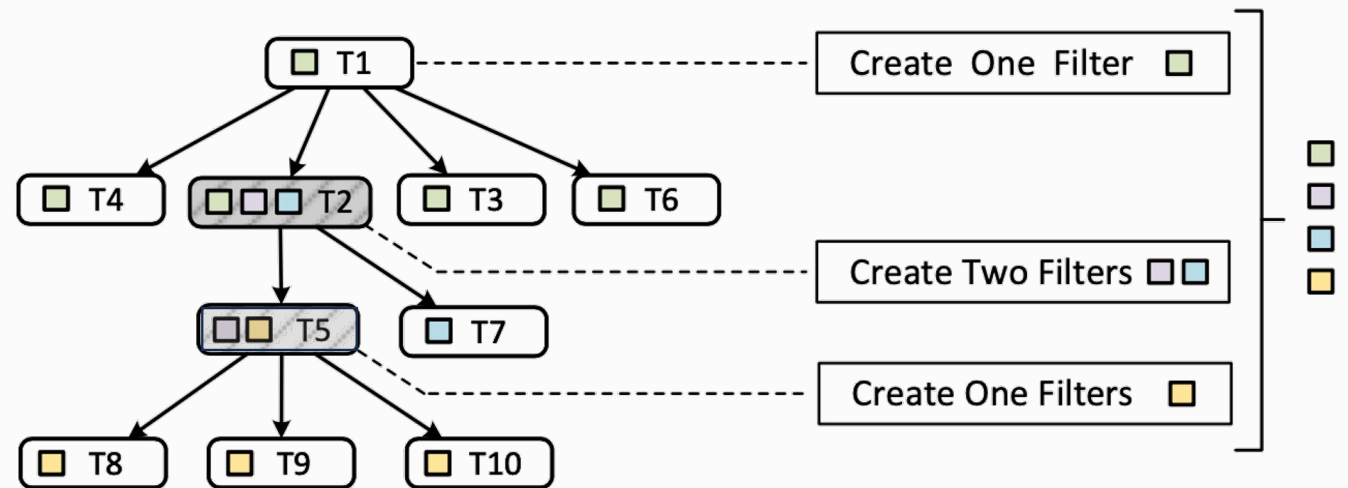
Asymmetric Transfer Plan

A deep, **chaining tree** is ideal for the forward pass, whereas a wide, **broadcast tree** is best for the backward pass.

□ Base Table □□ Bridge Table



Forward Pass: Chaining Tree

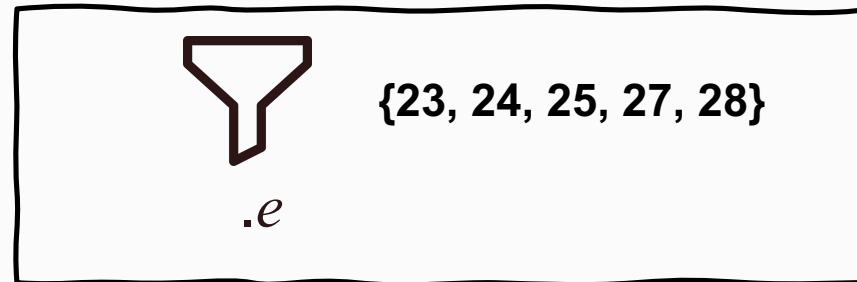


Backward Pass: Broadcast Tree

2. Storage-Format-Unaware Filtering

Bloom Filter cannot help skip row groups.

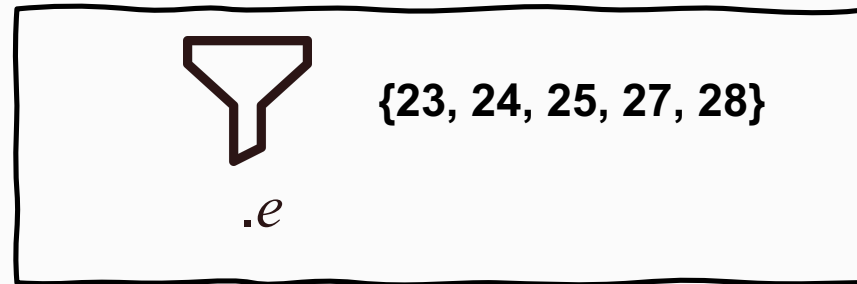
Row Group 1	a	c	e
	2	100	21
Row Group 2	4	100	22
	2	200	25
Row Group 3	1	100	26
	4	500	27
Row Group 4	1	300	31



2. Storage-Format-Unaware Filtering

Bloom Filter cannot help skip row groups.

Row Group 1	a	c	e
	2	100	21
Row Group 2	4	100	22
	2	200	25
Row Group 3	1	100	26
	4	500	27
Row Group 3	1	300	31

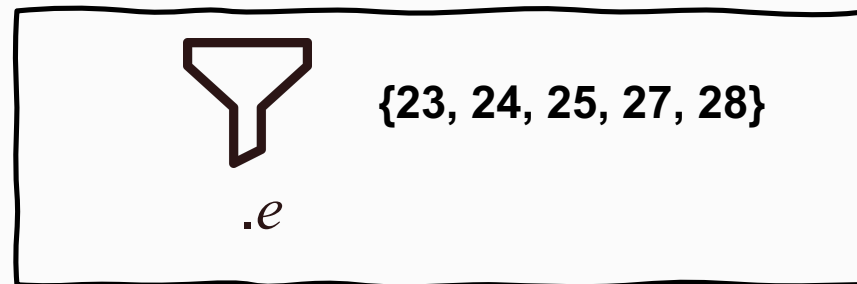


Stage 1: Data Scanning (#rows = 6)

2. Storage-Format-Unaware Filtering

Bloom Filter cannot help skip row groups.

	a	c	e
Row Group 1	2	100	21
	4	100	22
Row Group 2	2	200	25
	1	100	26
Row Group 3	4	500	27
	1	300	31



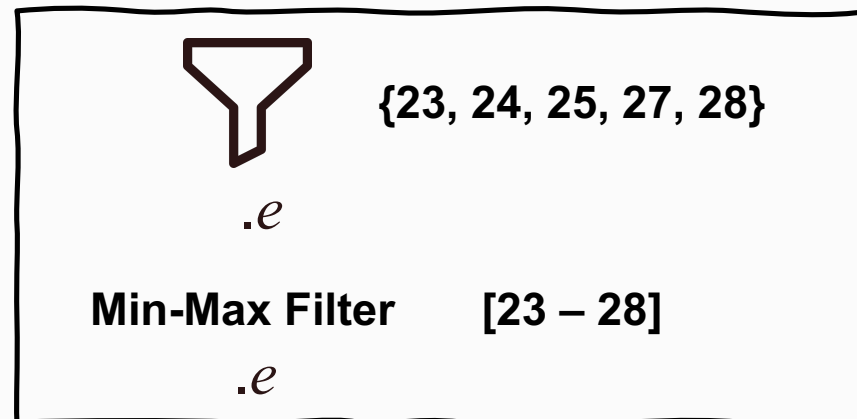
Stage 1: Data Scanning (#rows = 6)

Stage 2: Bloom Filter Probing (#rows = 6)

Cascade Filter

Cascade Filter = Min-max Filter + Bloom Filter + others (optional)

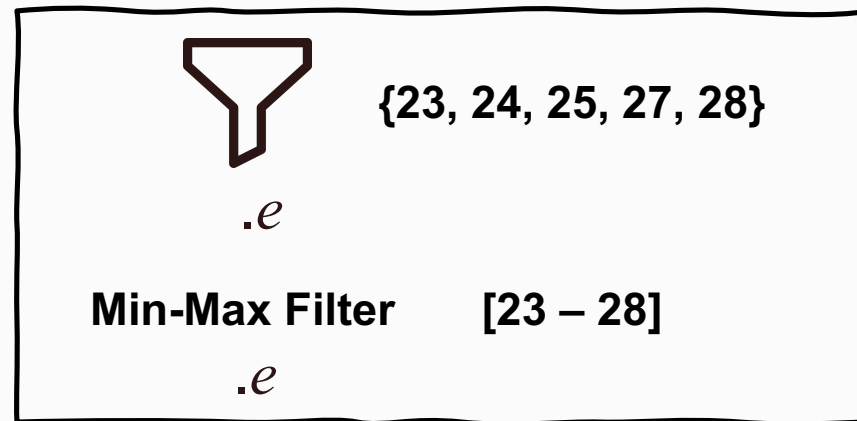
	a	c	e
Row Group 1	2	100	21
	4	100	22
Row Group 2	2	200	25
	1	100	26
Row Group 3	4	500	27
	1	300	31



Cascade Filter

Cascade Filter = Min-max Filter + Bloom Filter + others (optional)

	a	c	e
Row Group 1	2	100	21
	4	100	22
Row Group 2	2	200	25
	1	100	26
Row Group 3	4	500	27
	1	300	31

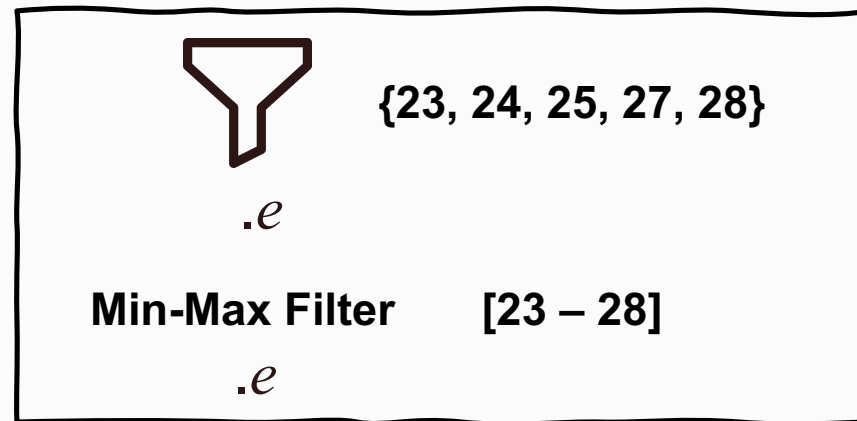


Stage 1: Data Scanning (#rows = 4)

Cascade Filter

Cascade Filter = Min-max Filter + Bloom Filter + others (optional)

	a	c	e
Row Group 1	2	100	21
	4	100	22
Row Group 2	2	200	25
	1	100	26
Row Group 3	4	500	27
	1	300	31



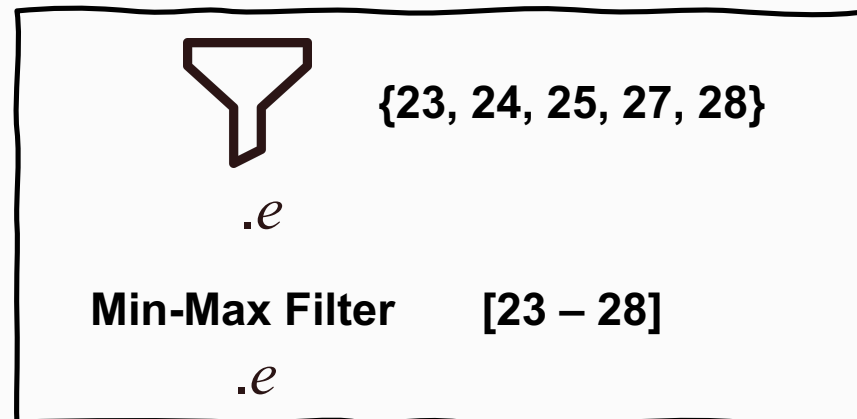
Stage 1: Data Scanning (#rows = 4)

Stage 2: Min-Max Probing (#rows = 4)

Cascade Filter

Cascade Filter = Min-max Filter + Bloom Filter + others (optional)

	a	c	e
Row Group 1	2	100	21
	4	100	22
Row Group 2	2	200	25
	1	100	26
Row Group 3	4	500	27
	1	300	31



Stage 1: Data Scanning (#rows = 4)

Stage 2: Min-Max Probing (#rows = 4)

Stage 3: Bloom Filter Probing (#rows = 3)

Non-Reducible Queries

If **all base table rows already contribute to the join results** before filtering, RPT causes regressions.

Non-Reducible Queries

If **all** base table rows already contribute to the join results before filtering, RPT causes regressions.

```
SELECT *  
FROM   R, S  
WHERE  R.a = S.a
```

a	f	g
1	xx	xx
1	xx	xx
3	xx	xx
1	xx	xx
2	xx	xx
5	xx	xx

R

a	b	c
1	10	100
2	20	100
3	10	300
1	40	300
5	10	100

Complete Domain



Dynamic Execution

Goal: collect “Information” from **efficiently filtered** tables only

A Table is efficiently filtered if and only if :

Filterable: It has at least one user-defined predicate.

Selective: Its predicates are selective.



Runtime Estimation



Transfer Plan Modification

Dynamic Execution

Goal: collect “Information” from **efficiently filtered** tables only

A Table is efficiently filtered if and only if :

Filterable: It has at least one user-defined predicate.

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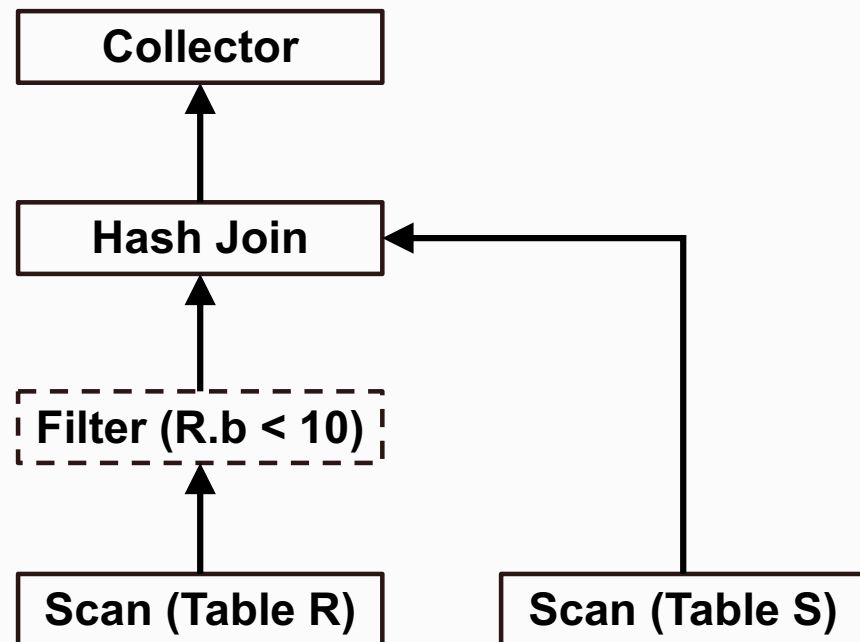
Runtime Estimation



Transfer Plan Modification
(refer to the paper)

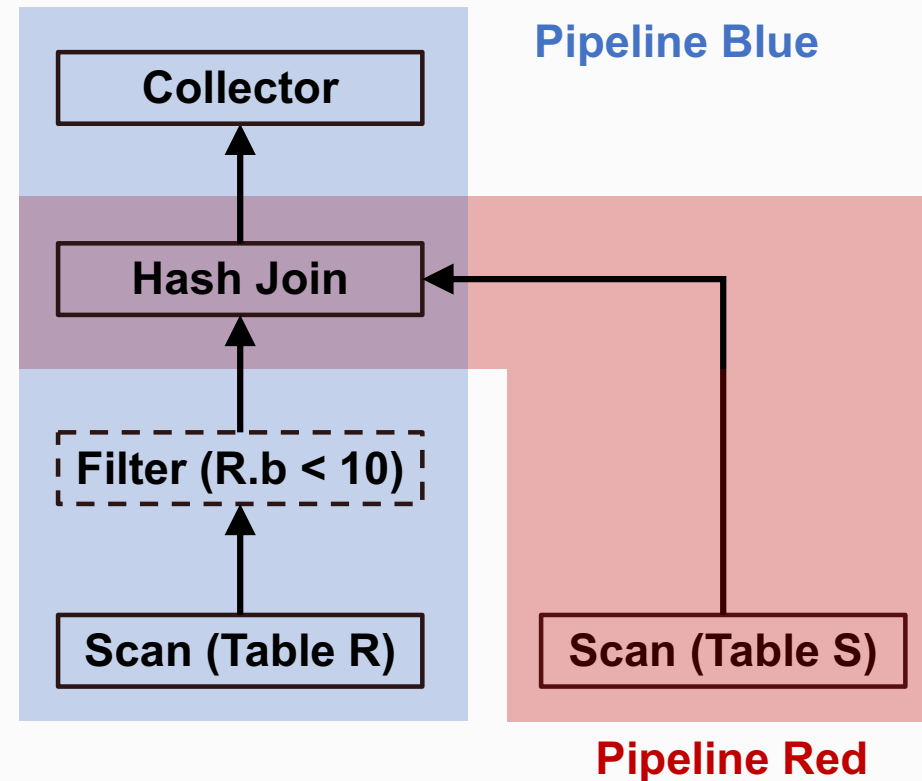
Estimate Selectivity at Runtime

```
SELECT *  
FROM   R, S  
WHERE  R.a = S.a  
AND    R.b < 10
```



Estimate Selectivity at Runtime

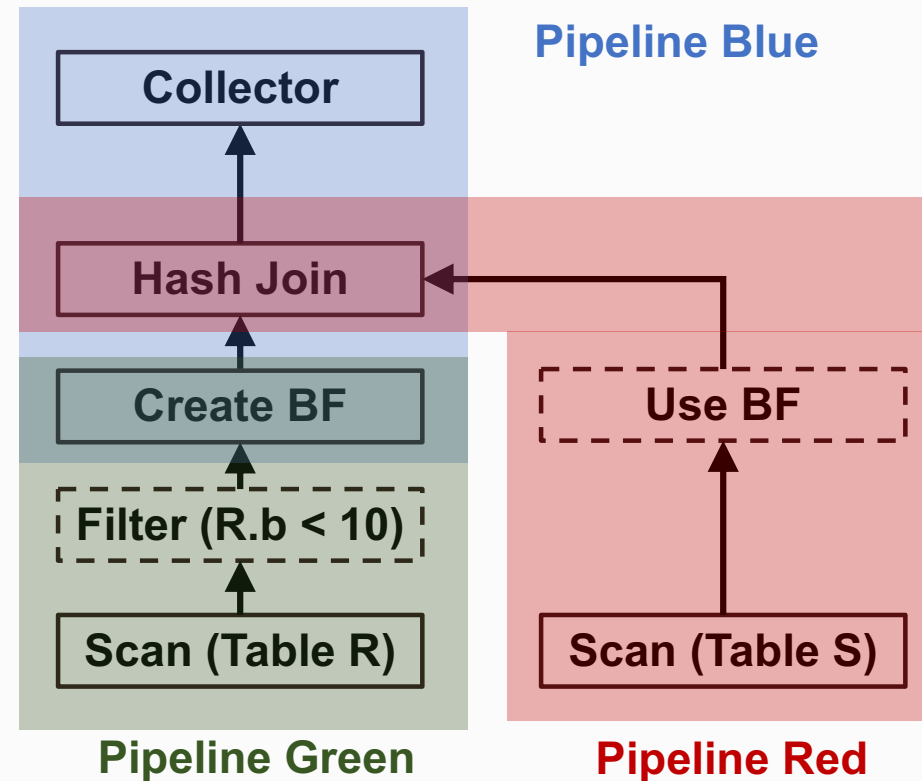
```
SELECT *  
FROM   R, S  
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```



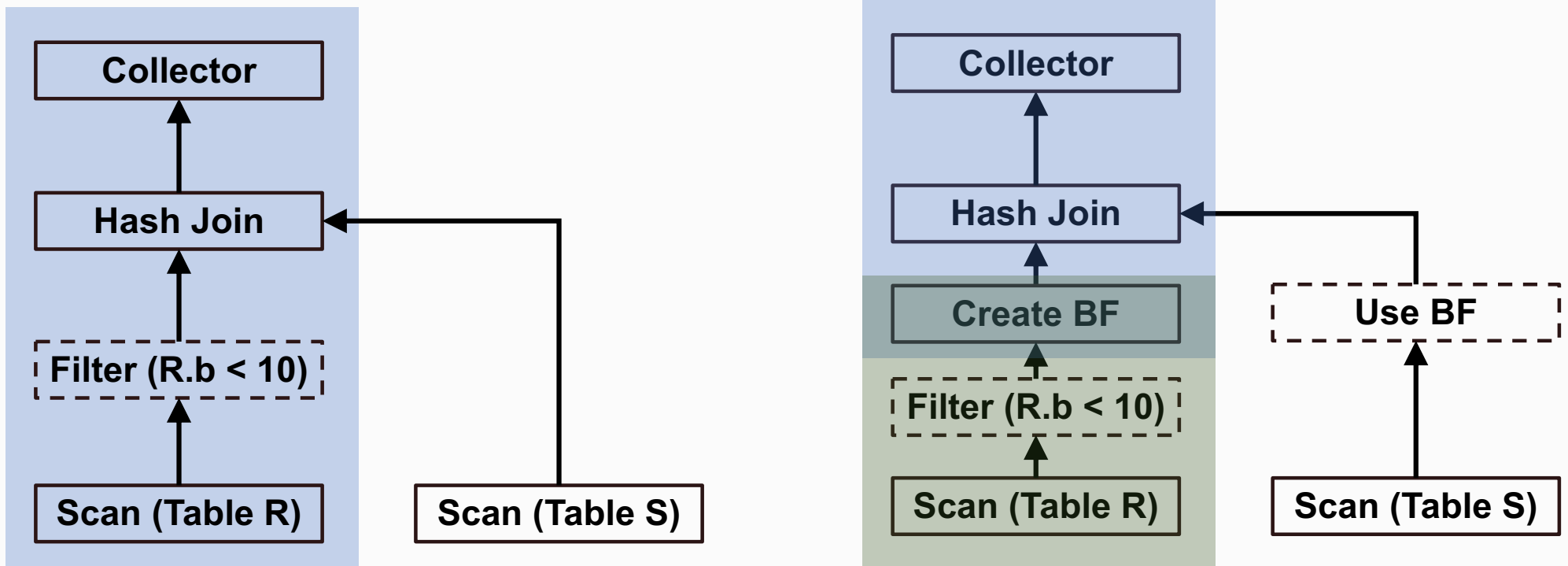
Estimate Selectivity at Runtime

```
SELECT *  
FROM   R, S  
WHERE  R.a = S.a  
AND    R.b < 10
```

To Create Runtime Filters, we **Split** the Pipeline



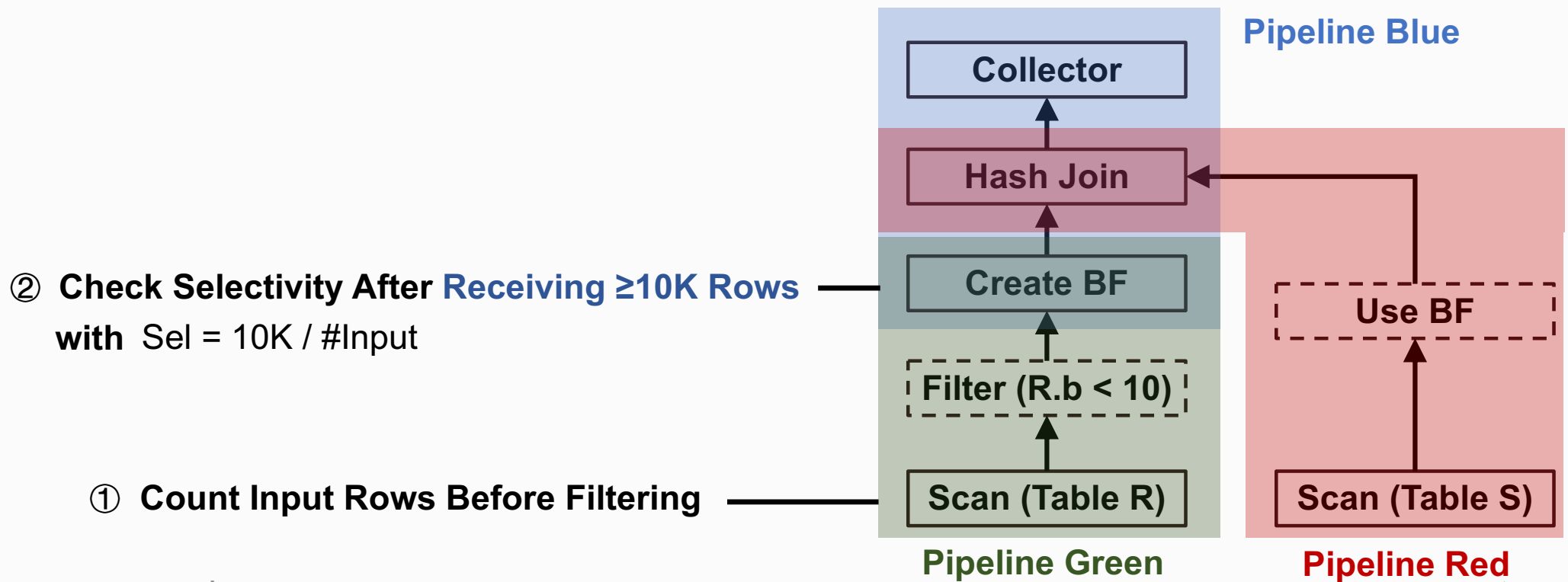
Estimate Selectivity at Runtime



One CreateBF = One Table Materlization

Estimate Selectivity at Runtime

Monitor **chunk statistics at runtime** to estimate selectivity

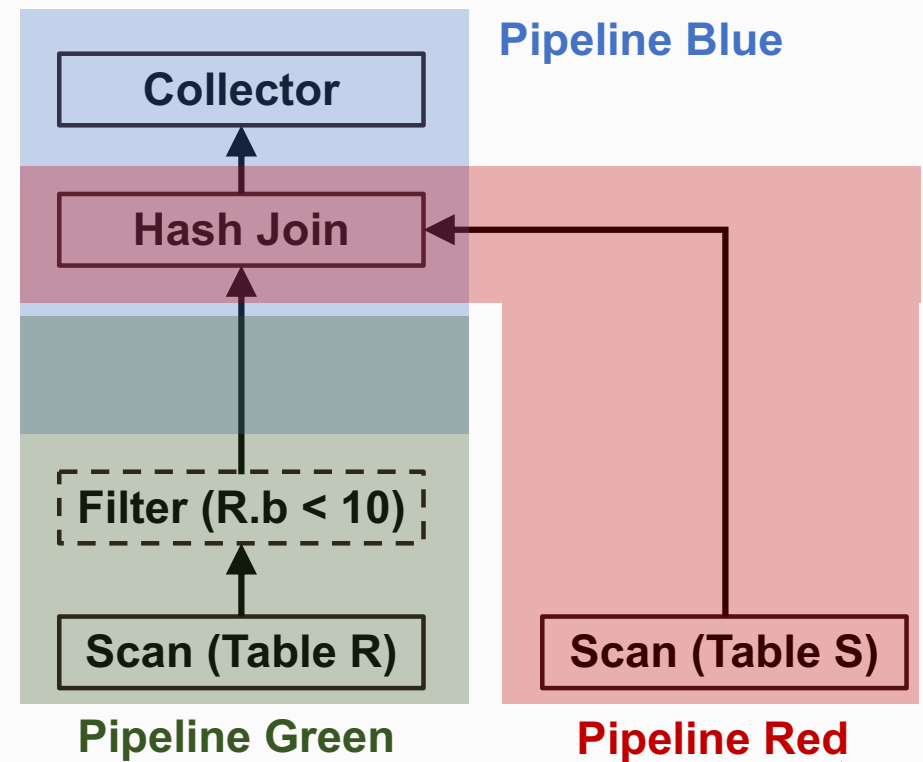


Estimate Selectivity at Runtime

Monitor **chunk statistics at runtime** to estimate selectivity

If the estimated selectivity is not desired,

1. **Cancel** BF construction

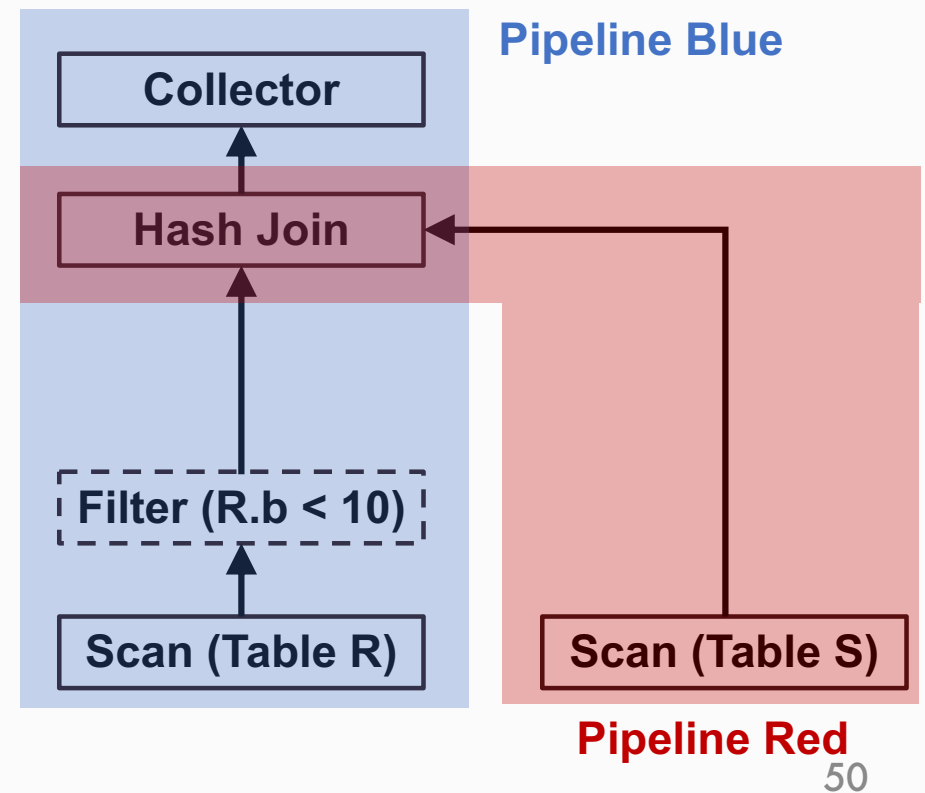


Estimate Selectivity at Runtime

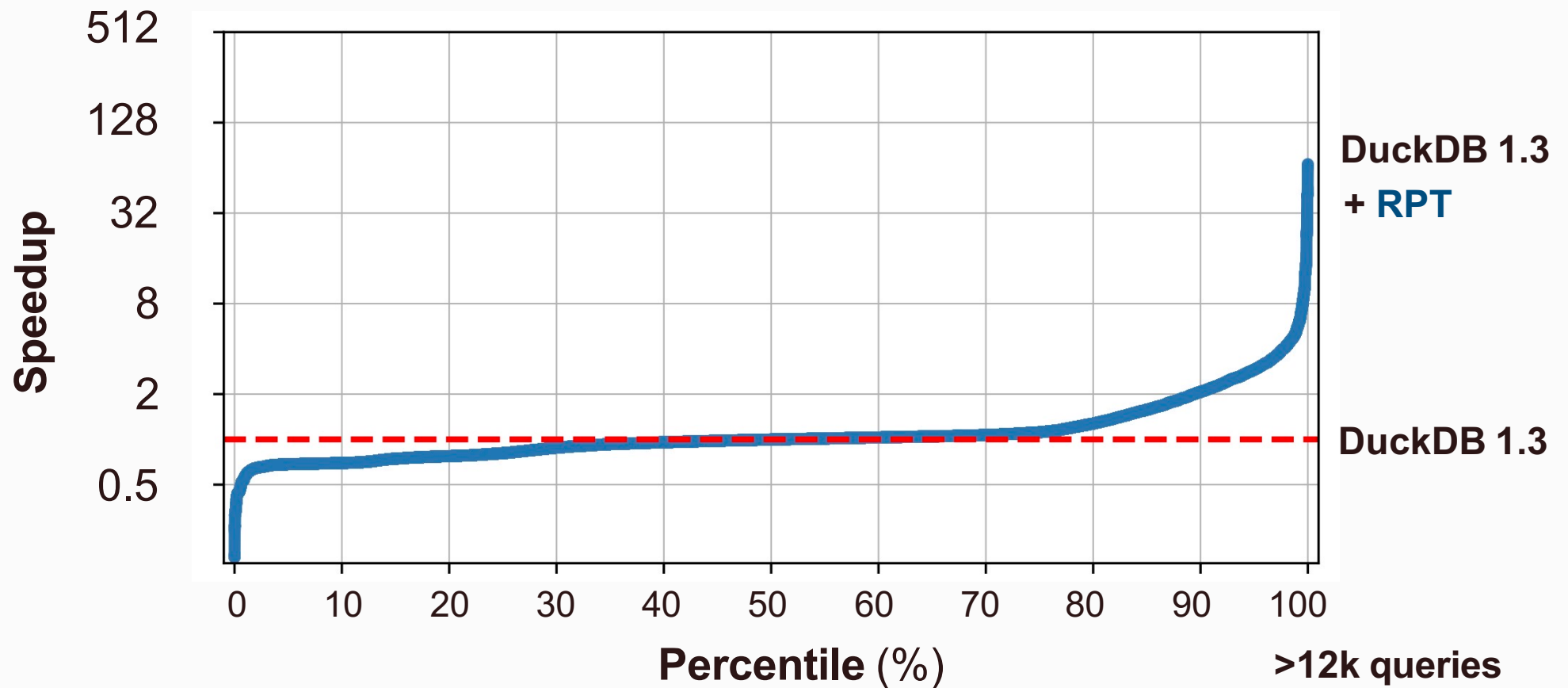
Monitor **chunk statistics at runtime** to estimate selectivity

If the estimated selectivity is not desired,

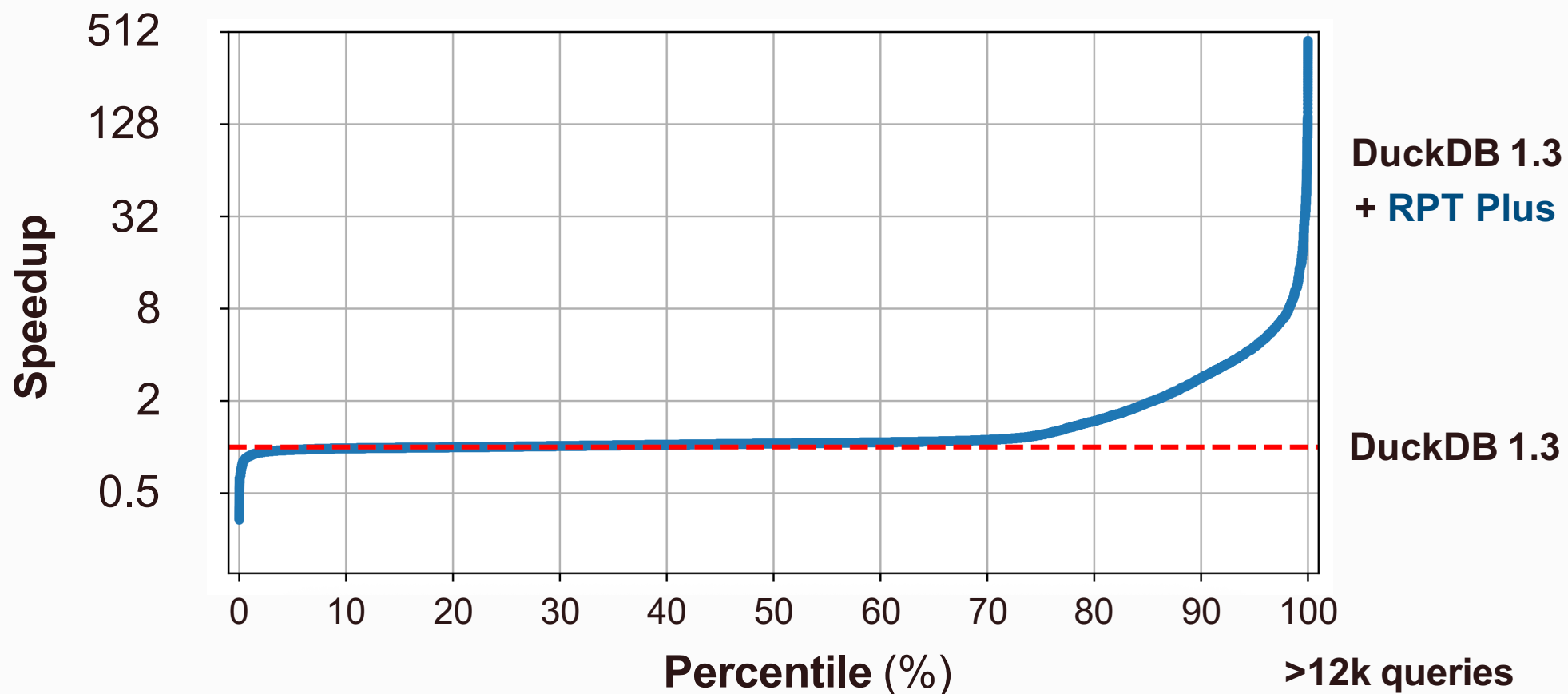
1. **Cancel** BF construction
2. **Move** operators across pipelines



SQLStorm: RPT vs. DuckDB 1.3



SQLStorm: RPT **Plus** vs. DuckDB 1.3



Takeaway

- Runtime filtering + data pruning = faster queries.
- Adapt at runtime, rather than trust cardinality estimates.
- Make predicate transfer a standalone execution layer.

Bloom: Beyond DuckDB

Towards Robust Join Acceleration for Every Execution Engine.



DuckDB Extension



Postgres Extension



Datafusion Extension

Same idea. Different engines. Robust speedup.