

Fine-grained Provenance for Linear Algebra Operators

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Motivation

- Provenance is well-understood for **relational data / queries**.
 - E.g., view maintenance, delete propagation, computing trust, prob. db
- But increasingly analysts are performing more complex tasks:
 - Machine learning, data mining, image analysis, graph analytics
- **Array data and matrix algebra** are commonly used!

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- **Array data** and **matrix algebra** are commonly used!
- Question: How do we **track provenance** in this setting?

Inspiration: Provenance Semirings

- An algebra framework [Green et al. PODS'07] for
 - **annotating** tuples in a relation
 - **propagating** annotations through relational queries (SPJU and aggregation)
- Enables efficient **delete propagation, view maintenance, etc**

Semiring example: input data

P

PID	PValue
101	11
102	12
2003	13
2004	14
2005	15
2006	16

Q

QID	QValue
2003	0
2004	0
2005	0
2006	0

R

RID	RValue
101	13
102	14
5005	15
5006	16
5007	17
5008	18
5009	19

Semiring example: query

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S(Value) :- P(x, Value), R(x, _)

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S(Value) :- R(_, Value)

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S(Value) :- P(x, Value), R(x, _)

S(Value) :- P(x, Value), Q(x, _)

S(Value) :- R(_, Value)

Semiring example: output tuples

S

Value
11
12
13
14
15
16
17
18
19

S(Value) :- P(x, Value), R(x, _)
S(Value) :- P(x, Value), Q(x, _)
S(Value) :- R(_, Value)

Semiring example: annotated output

S

Value	Annotation
11	pr
12	pr
13	$pq + r$
14	$pq + r$
15	$pq + r$
16	$pq + r$
17	r
18	r
19	r

S(Value) :- P(x, Value), R(x, _)

S(Value) :- P(x, Value), Q(x, _)

S(Value) :- R(_, Value)

Semiring example: delete propagation

S

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11	pr
12	pr
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15	$pq + r$
16	$pq + r$
17	r
18	r
19	r

What if we remove tuples from \mathbf{P} ?

Set $p = 0$!

Semiring example: delete propagation

S

Value	Annotation
11	0
12	0
13	r
14	r
15	r
16	r
17	r
18	r
19	r

What if we remove tuples from **P**?

Set $p = 0$!

Semimodule example: aggregation

S

Query: $\text{SUM}(\text{Value})$

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12	pr
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Semimodule example: annotation

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Query: **SUM(Value)**

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18	r
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Annotated aggregation

$$\begin{aligned}
 & pr * (11 + 12) + (pq + r) * (13 + 14 + 15 + 16) \\
 & + r * (17 + 18 + 19)
 \end{aligned}$$

Semimodule example: annotation

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Query: $\text{SUM}(\text{Value})$

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13	$pq + r$
14	$pq + r$
15	$pq + r$
16	$pq + r$
17	r
18	r
19	r

Annotated aggregation

$$\begin{aligned} & \textcolor{red}{pr} * (11 + 12) + (\textcolor{red}{pq + r}) * (13 + 14 + 15 + 16) \\ & + \textcolor{red}{r} * (17 + 18 + 19) \end{aligned}$$

First term: annotation

Semimodule example: annotation

S

Query: $\text{SUM}(\text{Value})$

Value	Annotation
11	pr
12	pr
13	$pq + r$
14	$pq + r$
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19	r

Annotated aggregation

$$pr * (11 + 12) + (pq + r) * (13 + 14 + 15 + 16) \\ + r * (17 + 18 + 19)$$

First term: annotation

Second term: value

Semimodule example: annotation

S

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13	$pq + r$
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19	r

Annotated aggregation

$$pr * (11 + 12) + (pq + r) * (13 + 14 + 15 + 16) \\ + r * (17 + 18 + 19)$$

or $pr * 23 + (pq + r) * 58 + r * 54$

Semimodule example: annotation

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Query: **SUM(Value)**

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or $pr * 23 + (pq + r) * 58 + r * 54$

or $pq * 58 + r * (p * 23 + 112)$

Semimodule example: annotation

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Query: **SUM(Value)**

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Annotated aggregation

$$pr * (11 + 12) + (pq + r) * (13 + 14 + 15 + 16) \\ + r * (17 + 18 + 19)$$

or $pr * 23 + (pq + r) * 58 + r * 54$

or $pq * 58 + r * (p * 23 + 112)$

Delete propagation:

set $r = 0$ and obtain $pq * 58$

Tracking matrix provenance

- We want to get the same benefits!
 - Algebraically manipulate annotated matrices
 - Hypothetical deletion

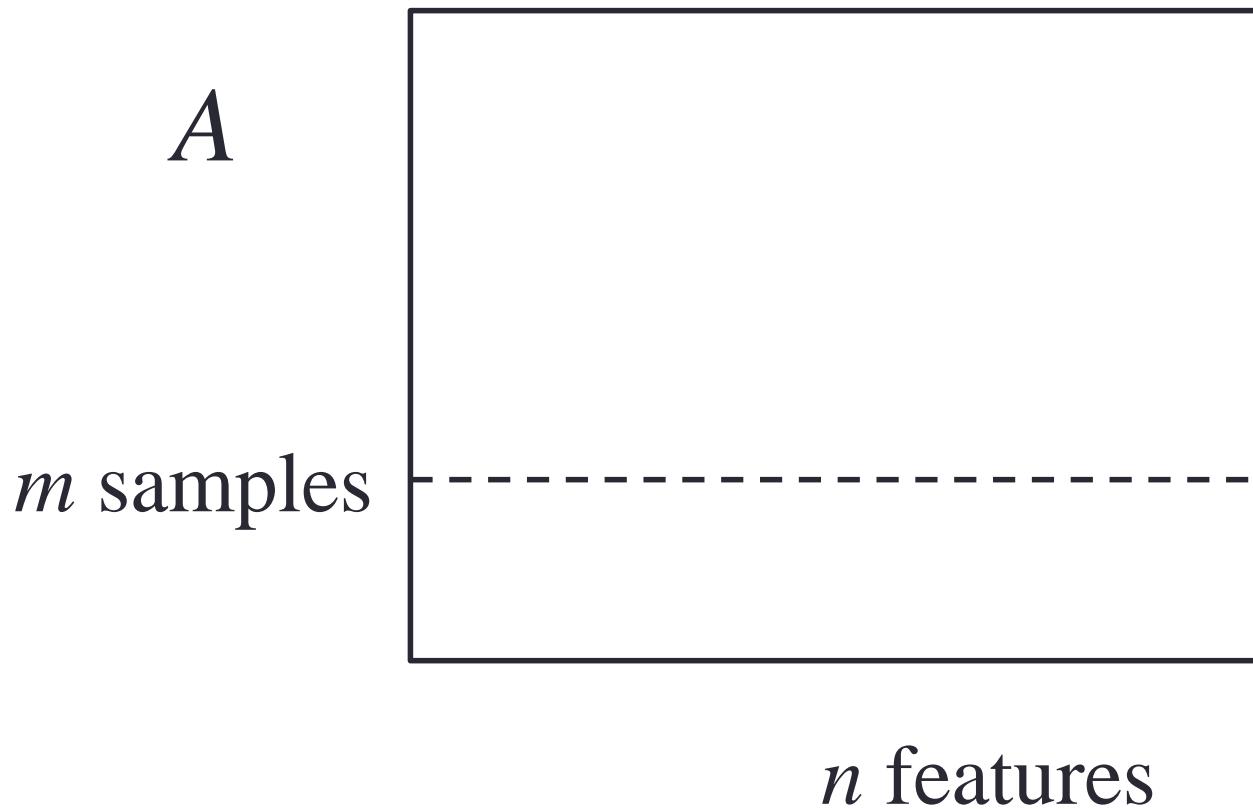
Tracking matrix data

A
 m samples

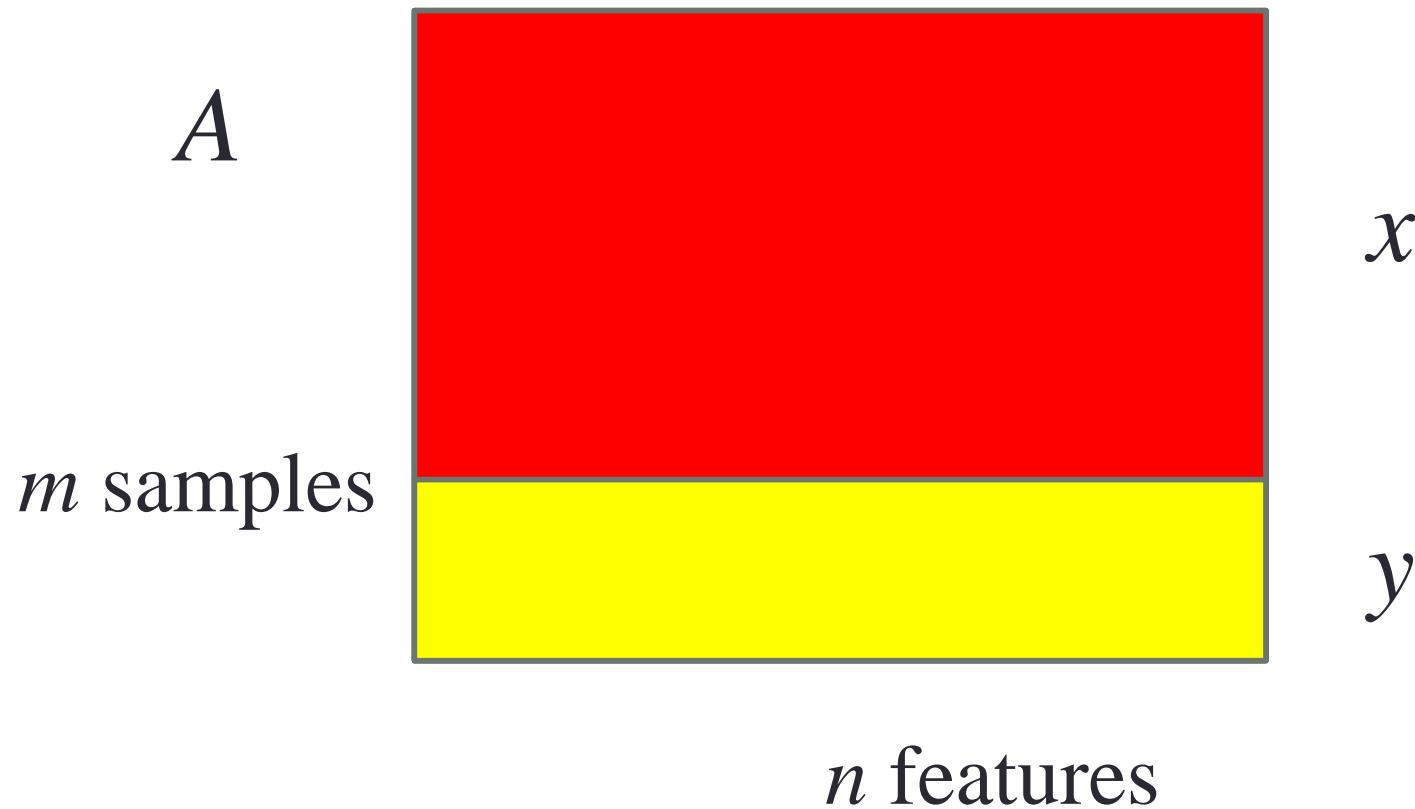


n features

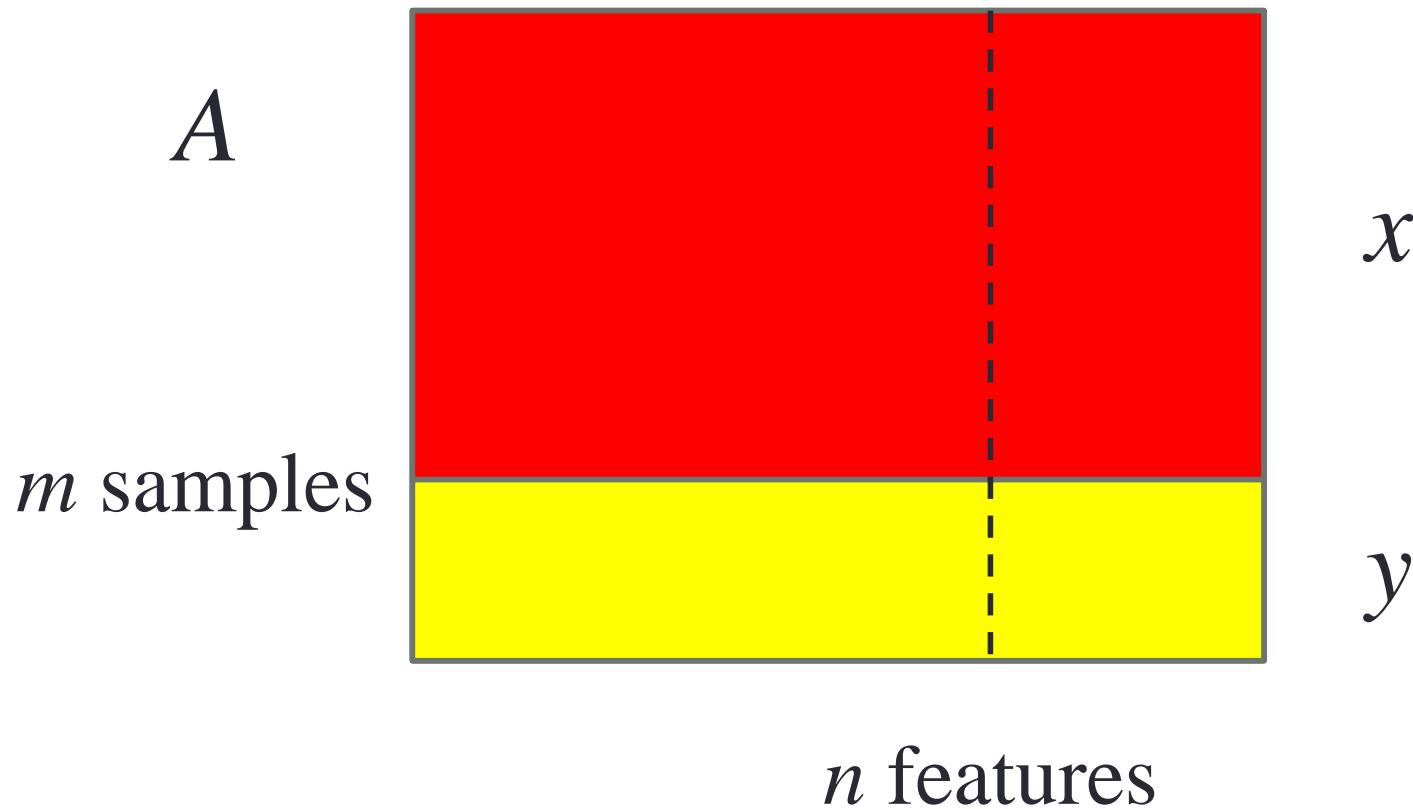
Tracking matrix data: partitioning



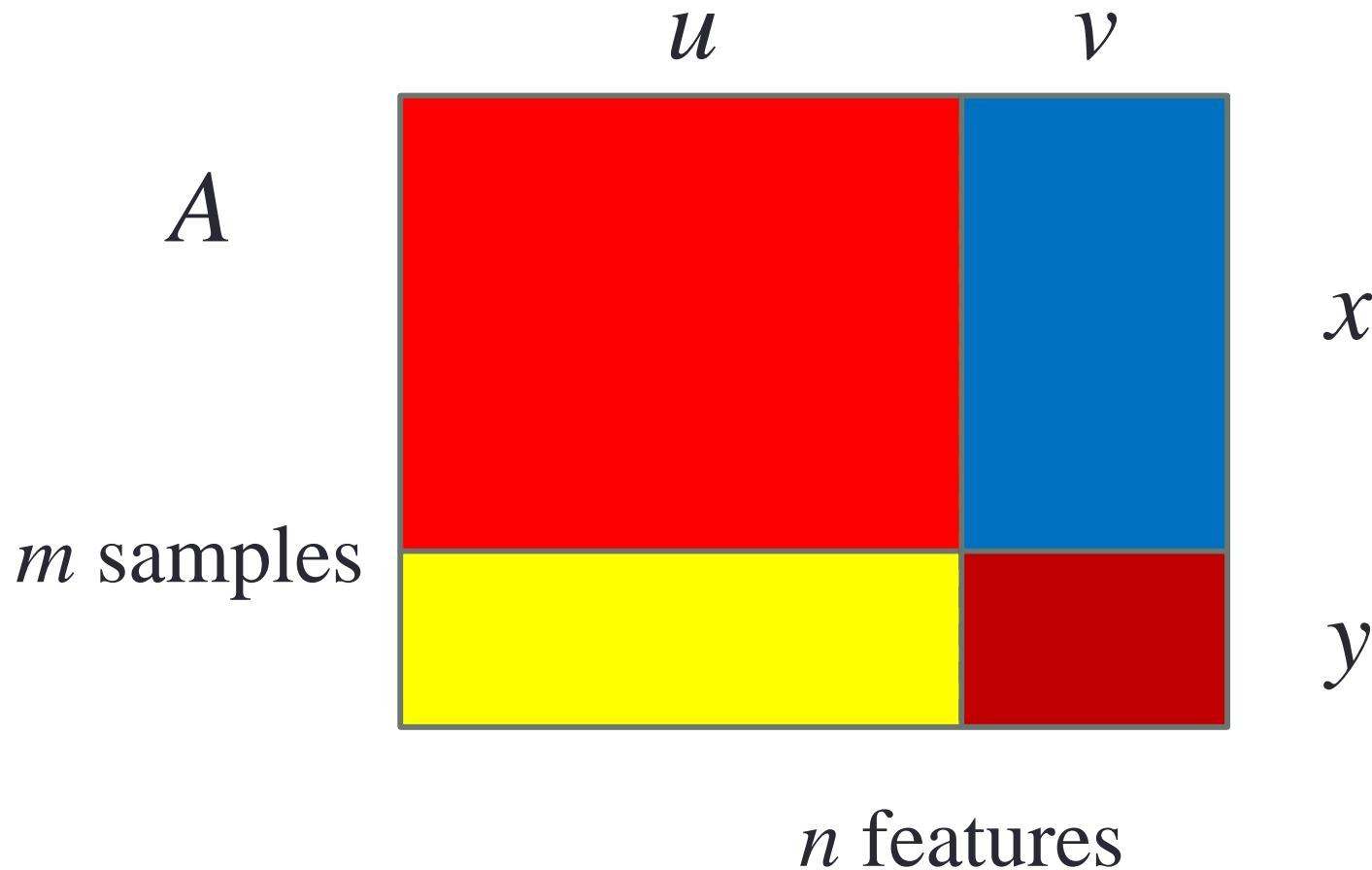
Tracking matrix data: annotating



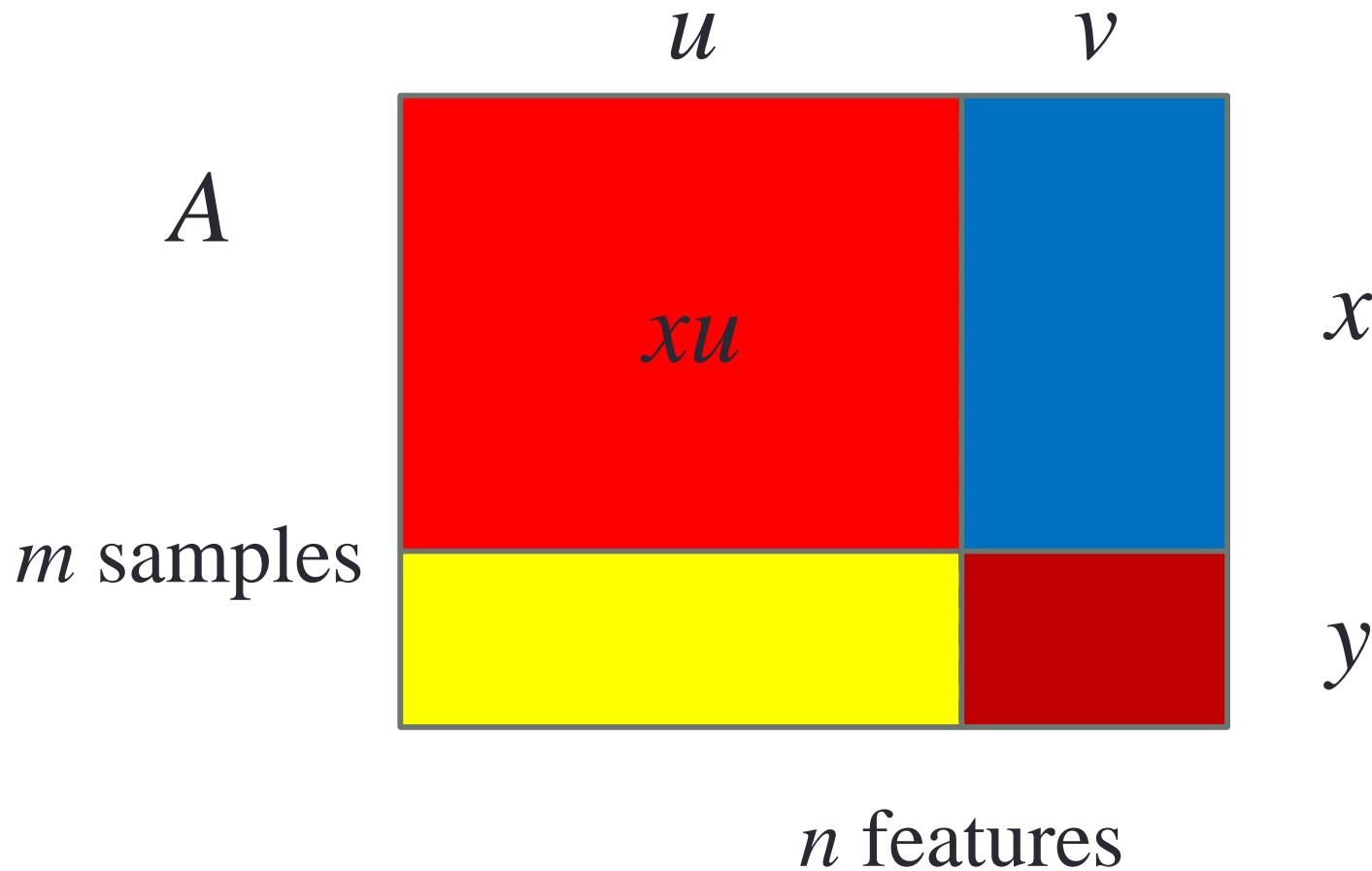
Tracking matrix data: annotating



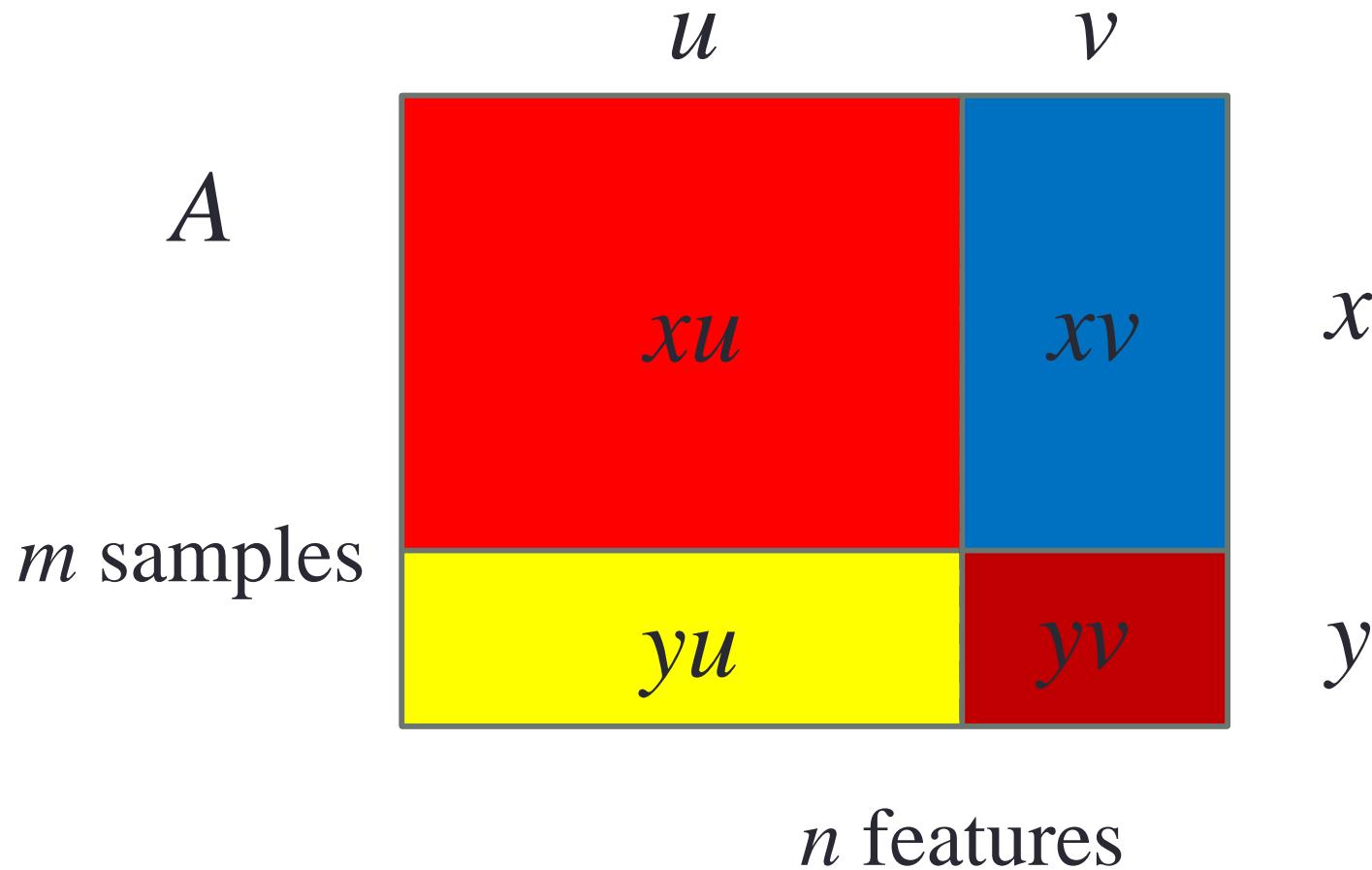
Tracking matrix data: annotating



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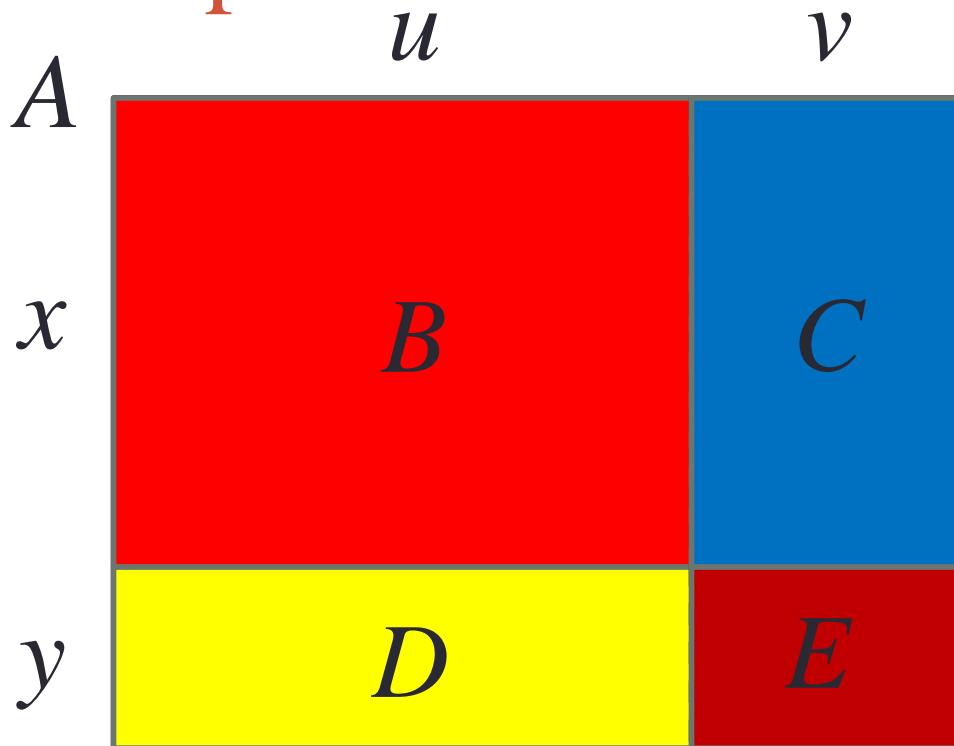
Tracking matrix data: annotating



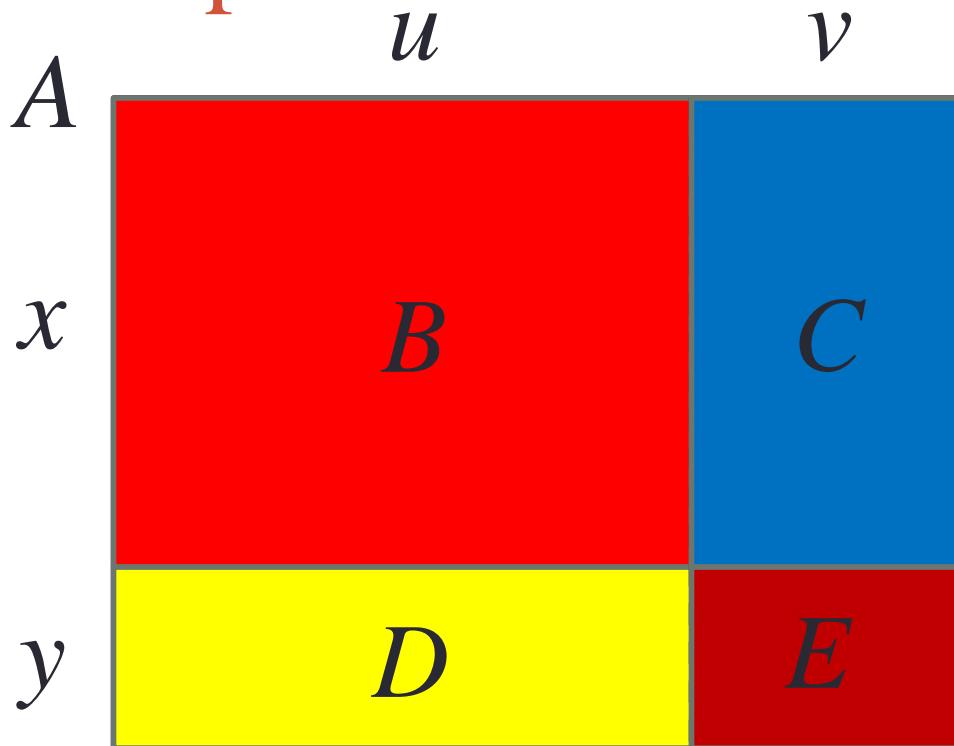
Challenges

- Specify and relate different parts of a given matrix
 - Matrix decomposition through **selector matrices**
- Specify connection between derived and source matrices
 - Embed matrix algebra and provenance into a **semialgebra**

Decomposition



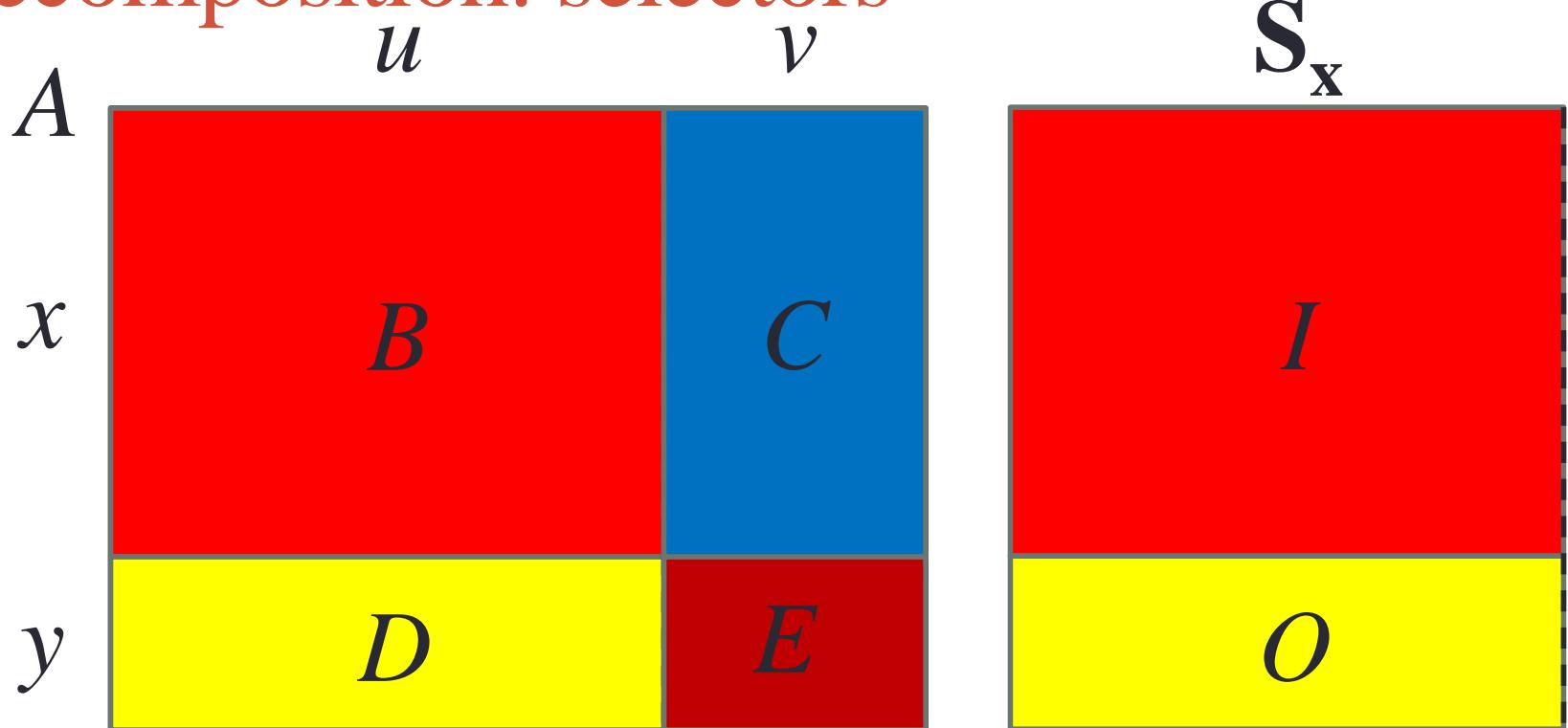
Decomposition: selectors



$$A = \mathbf{S}_x \mathbf{B} \mathbf{T}_u + \mathbf{S}_x \mathbf{C} \mathbf{T}_v + \mathbf{S}_y \mathbf{D} \mathbf{T}_u + \mathbf{S}_y \mathbf{E} \mathbf{T}_v$$

with selectors $\mathbf{S}_x \mathbf{S}_y \mathbf{T}_u \mathbf{T}_v$

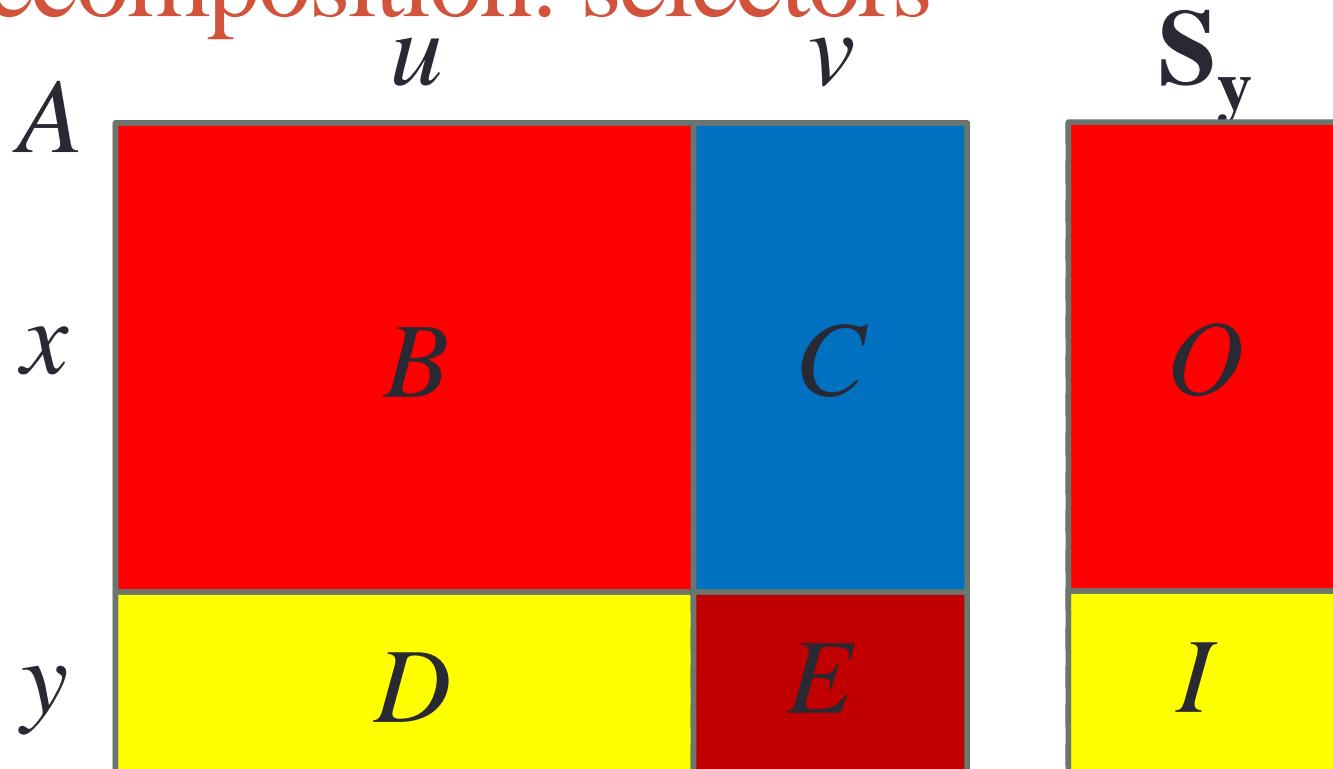
Decomposition: selectors



$$A = S_x \textcolor{red}{B} T_u + S_x \textcolor{blue}{C} T_v + S_y \textcolor{brown}{D} T_u + S_y \textcolor{red}{E} T_v$$

with selectors $S_x S_y T_u T_v$

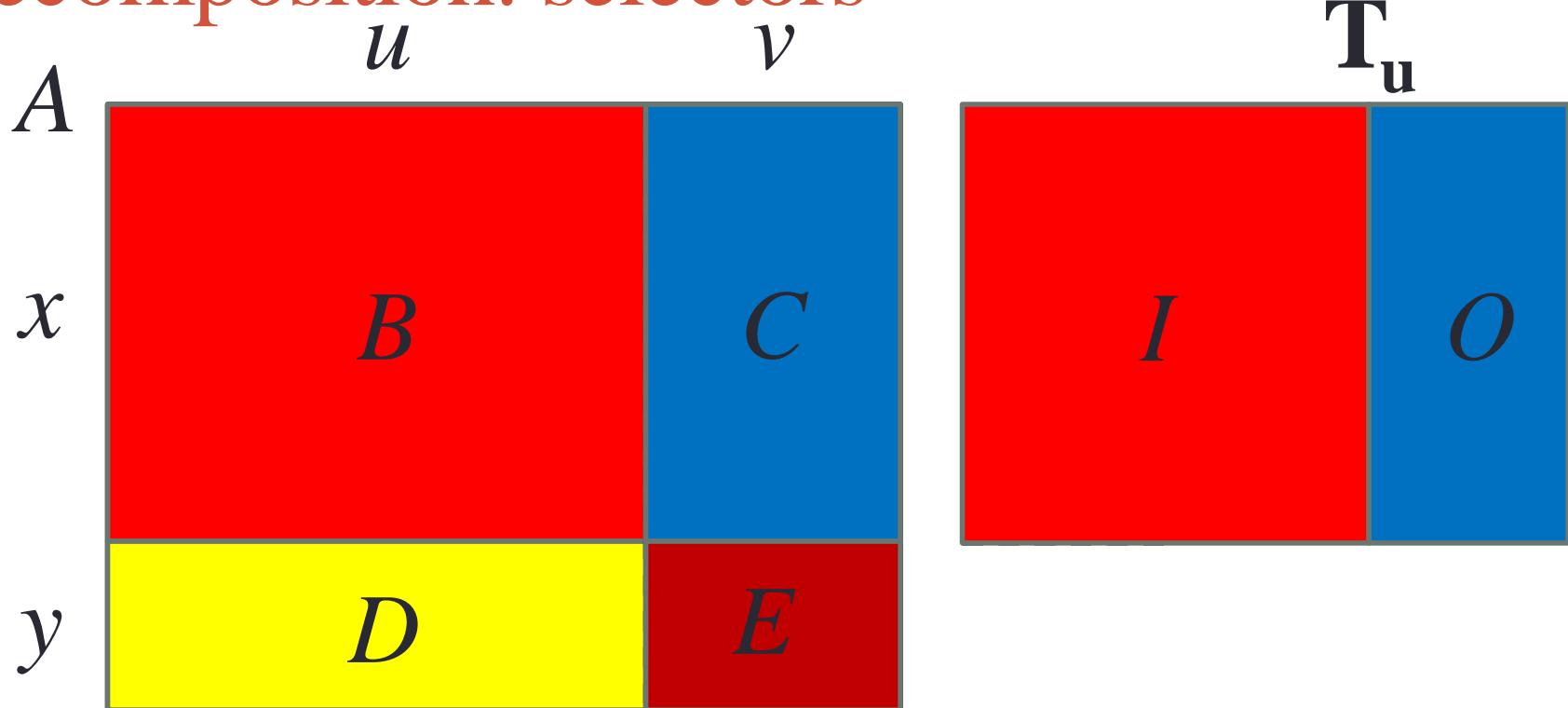
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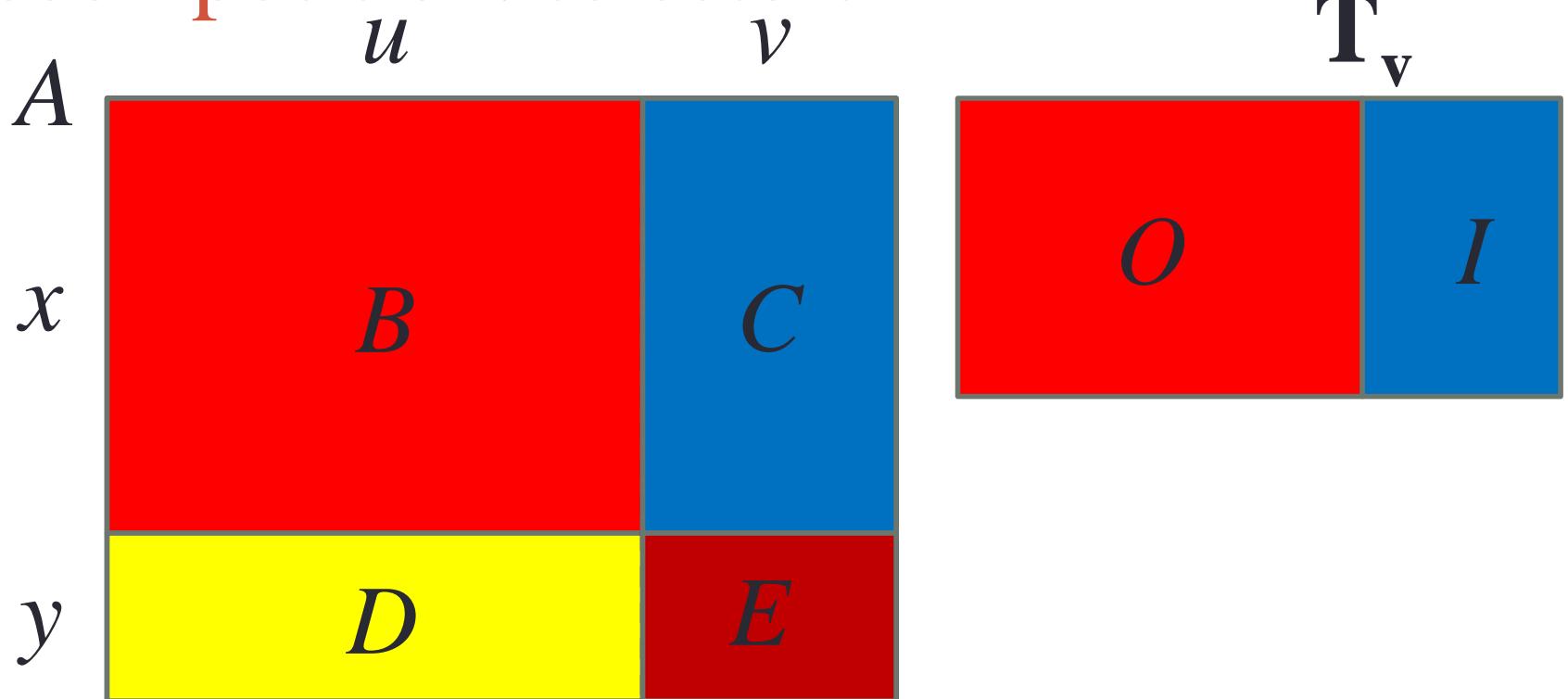
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Decomposition: selectors



$$A = \mathbf{S}_x \mathbf{B} \mathbf{T}_u + \mathbf{S}_x \mathbf{C} \mathbf{T}_v + \mathbf{S}_y \mathbf{D} \mathbf{T}_u + \mathbf{S}_y \mathbf{E} \mathbf{T}_v$$

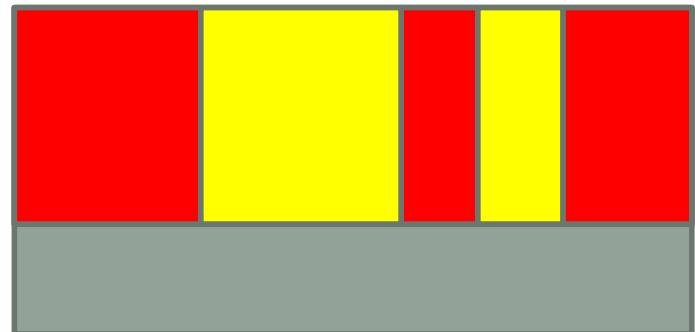
with selectors $\mathbf{S}_x \mathbf{S}_y \mathbf{T}_u \mathbf{T}_v$

Summary: selectors

- Relate a matrix to its sub-matrices.
- Matrices with only 0/1.
- Any row / column has at most a 1.

Summary: selectors

- Relate a matrix to its sub-matrices.
 - Matrices with only 0/1.
 - Any row / column has at most a 1.
-
- Extends to **non-adjacent** case.
 - Works for any **rectangular partition**.



Provenance propagation

- We have
 - Matrices and operators over them – **Algebra of matrices** \mathcal{M}
 - Annotations – **Semiring of provenance polynomials** $\mathbb{N}[X]$
- Goals
 - Combine annotations in the same structure as the matrices
 - Operations should **propagate** data and annotations

Provenance propagation

- We have
 - Matrices and operators over them – **Algebra of matrices** \mathcal{M}
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- Goals
 - Combine annotations in the same structure as the matrices
 - Operations should **propagate** data and annotations
- We do this in the space of tensor product $\mathbb{N}[X] \otimes \mathcal{M}$
 - Matrices as vectors, provenance as scalars: $p * A$
 - Satisfies all the laws of a **$\mathbb{N}[X]$ -semialgebra**

Laws of a $\mathbf{N}[X]$ -semialgebra (K -semialgebra)

$(K, +_K, \cdot_K, 0_K, 1_K)$ commutative semiring

$(K \otimes \mathcal{M}, +, \cdot, 0, I)$ forms a ring (just like the matrices)

laws for scalar multiplication in a K -semialgebra

$$k * (A_1 + A_2) = k * A_1 + k * A_2$$

$$k * 0 = 0$$

$$(k_1 +_K k_2) * A = k_1 * A + k_2 * A$$

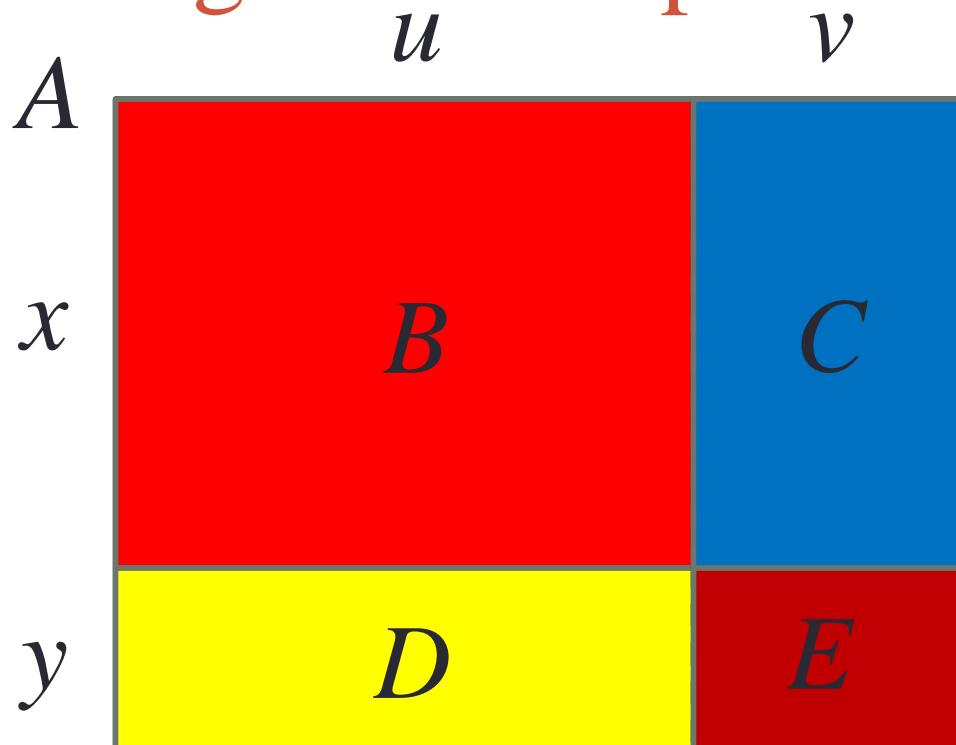
$$0_K * A = 0$$

$$(k_1 \cdot_K k_2) * A = k_1 * (k_2 * A)$$

$$1_K * A = A$$

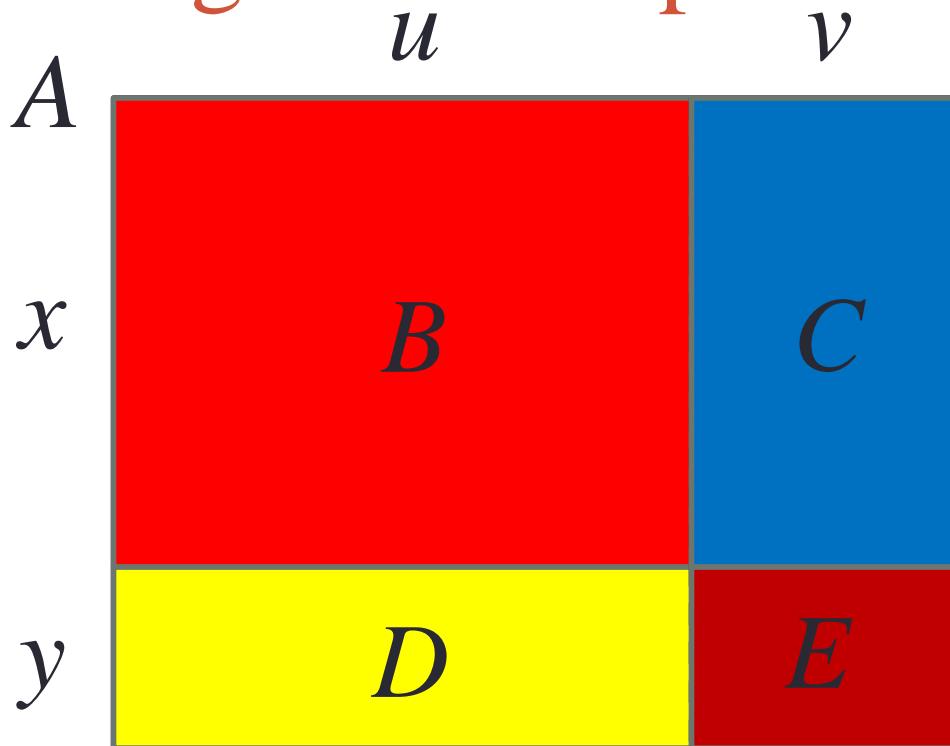
$$(k_1 * A_1)(k_2 * A_2) = (k_1 \cdot_K k_2) * (A_1 A_2)$$

Semialgebra example



$$A = \mathbf{S}_x \mathbf{B} \mathbf{T}_u + \mathbf{S}_x \mathbf{C} \mathbf{T}_v + \mathbf{S}_y \mathbf{D} \mathbf{T}_u + \mathbf{S}_y \mathbf{E} \mathbf{T}_v$$

Semialgebra example: add annotation



$$\begin{aligned} A = & \mathbf{S}_x \textcolor{red}{xu^*B} \mathbf{T}_u + \mathbf{S}_x \textcolor{blue}{xv^*C} \mathbf{T}_v \\ & + \mathbf{S}_y \textcolor{orange}{yu^*D} \mathbf{T}_u + \mathbf{S}_y \textcolor{red}{vy^*E} \mathbf{T}_v \end{aligned}$$

Semialgebra example: propagate annotation

$$\begin{aligned} A &= \mathbf{S}_x \textcolor{red}{xu*B T_u} + \mathbf{S}_x \textcolor{blue}{xv*C T_v} \\ &+ \mathbf{S}_y \textcolor{yellow}{yu*D T_u} + \mathbf{S}_y \textcolor{red}{vy*E T_v} \end{aligned}$$

Propagating annotation: transposition

$$\begin{aligned} A = & \mathbf{S}_x \textcolor{red}{xu} * B \mathbf{T}_u + \mathbf{S}_x \textcolor{blue}{xv} * C \mathbf{T}_v \\ & + \mathbf{S}_y \textcolor{yellow}{yu} * D \mathbf{T}_u + \mathbf{S}_y \textcolor{red}{vy} * E \mathbf{T}_v \end{aligned}$$

$$\begin{aligned} A^T = & \mathsf{T}_u^T (xu * B^T) S_x^T + \mathsf{T}_u^T (yu * D^T) S_y^T \\ & + \mathsf{T}_v^T (xv * C^T) S_x^T + \mathsf{T}_v^T (yv * E^T) S_y^T \end{aligned}$$

Propagating annotation: transposition

$$\begin{aligned} A = & \mathbf{S}_x \textcolor{red}{xu^*B} \mathbf{T}_{\mathbf{u}} + \mathbf{S}_x \textcolor{blue}{xv^*C} \mathbf{T}_{\mathbf{v}} \\ & + \mathbf{S}_y \textcolor{yellow}{yu^*D} \mathbf{T}_{\mathbf{u}} + \mathbf{S}_y \textcolor{red}{vy^*E} \mathbf{T}_{\mathbf{v}} \end{aligned}$$

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Transposition of a selector is still a selector
 Still a sum of (selector \times matrix \times selector)

Propagating annotation: multiplication

$$\begin{aligned} A &= \mathbf{S}_x \textcolor{red}{xu * B} \mathbf{T}_{\mathbf{u}} + \mathbf{S}_x \textcolor{blue}{xv * C} \mathbf{T}_{\mathbf{v}} \\ &+ \mathbf{S}_y \textcolor{yellow}{yu * D} \mathbf{T}_{\mathbf{u}} + \mathbf{S}_y \textcolor{red}{vy * E} \mathbf{T}_{\mathbf{v}} \end{aligned}$$

$$\begin{aligned} A^T &= \mathbf{T}_u^T (xu * B^T) \mathbf{S}_x^T + \mathbf{T}_u^T (yu * D^T) \mathbf{S}_y^T \\ &+ \mathbf{T}_v^T (xv * C^T) \mathbf{S}_x^T + \mathbf{T}_v^T (yv * E^T) \mathbf{S}_y^T \end{aligned}$$

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$$\begin{aligned} AA^T = & \mathbf{S}_x (x^2 u^2 * BB^T + x^2 v^2 * CC^T) \mathbf{S}_x^T \\ & + \mathbf{S}_x (xyu^2 * BD^T + xyv^2 * CE^T) \mathbf{S}_y^T \\ & + \mathbf{S}_y (xyu^2 * DB^T + xyv^2 * EC^T) \mathbf{S}_x^T \\ & + \mathbf{S}_y (y^2 u^2 * DD^T + y^2 v^2 * EE^T) \mathbf{S}_y^T \end{aligned}$$

Propagating annotation: multiplication

$$\begin{aligned}
 A &= \mathbf{S}_x \textcolor{red}{xu * B} \mathbf{T}_u + \mathbf{S}_x \textcolor{blue}{xv * C} \mathbf{T}_v \\
 &+ \mathbf{S}_y \textcolor{yellow}{yu * D} \mathbf{T}_u + \mathbf{S}_y \textcolor{red}{vy * E} \mathbf{T}_v
 \end{aligned}
 \quad
 \begin{aligned}
 \mathbf{T}_u \mathbf{T}_u^T &= \mathbf{T}_v \mathbf{T}_v^T = \mathbf{I} \\
 \mathbf{T}_u \mathbf{T}_v^T &= \mathbf{T}_v \mathbf{T}_u^T = \mathbf{0}
 \end{aligned}$$

$$\begin{aligned}
 A^T &= \mathbf{T}_u^T (xu * B^T) \mathbf{S}_x^T + \mathbf{T}_u^T (yu * D^T) \mathbf{S}_y^T \\
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 \end{aligned}$$

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 AA^T &= \mathbf{S}_x (x^2 u^2 * BB^T + x^2 v^2 * CC^T) \mathbf{S}_x^T \\
 &+ \mathbf{S}_x (xyu^2 * BD^T + xyv^2 * CE^T) \mathbf{S}_y^T \\
 &+ \mathbf{S}_y (xyu^2 * DB^T + xyv^2 * EC^T) \mathbf{S}_x^T \\
 &+ \mathbf{S}_y (y^2 u^2 * DD^T + y^2 v^2 * EE^T) \mathbf{S}_y^T
 \end{aligned}$$

Semialgebra: delete propagation

$$\begin{aligned} AA^T &= S_x(x^2u^2 * BB^T + x^2v^2 * CC^T)S_x^T \\ &+ S_x(xyu^2 * BD^T + xyv^2 * CE^T)S_y^T \\ &+ S_y(xyu^2 * DB^T + xyv^2 * EC^T)S_x^T \\ &+ S_y(y^2u^2 * DD^T + y^2v^2 * EE^T)S_y^T \end{aligned}$$

Semialgebra: delete propagation

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 &+ S_y(xyu^2 * DB^T + xyv^2 * EC^T)S_x^T \\
 &+ S_y(y^2 u^2 * DD^T + y^2 v^2 * EE^T)S_y^T
 \end{aligned}$$

Deletion propagation: set $y = 0$

$$AA^T = x^2 * (u^2 * S_x BB^T S_x^T + v^2 * S_x CC^T S_x^T)$$

Preliminary application: solving equations

- $(A + B)x = b$, A and B are square matrices
- A is from source p , B is from source q

Preliminary application: solving equations

- $(A + B)x = b$, A and B are square matrices
- A is from source p , B is from source q
- Jacobi method: iteratively compute

$$u_{k+1} = (M^{-1}N)u_k + M^{-1}b \quad u_0 = \bar{0}$$

- $M = p * \text{diag}(A)$, $N = p * (\text{diag}(A) - A) - q * B$

Jacobi method: example

- Iteratively compute

$$u_{k+1} = (M^{-1}N)u_k + M^{-1}b \quad u_0 = \bar{0}$$

- $M = p * \text{diag}(A)$, $N = p * (\text{diag}(A) - A) - q * B$

Jacobi method: example

- Iteratively compute

$$u_{k+1} = (M^{-1}N)u_k + M^{-1}b \quad u_0 = \bar{0}$$

- $M = p * \text{diag}(A)$, $N = p * (\text{diag}(A) - A) - q * B$

$$A = p * \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}, B = q * \begin{bmatrix} 0 & 0 \\ -2 & 0 \end{bmatrix}, b = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$u_1 = p * \begin{bmatrix} \frac{1}{2} \\ -\frac{1}{2} \end{bmatrix}, u_2 = p * \begin{bmatrix} \frac{1}{2} \\ -\frac{1}{2} \end{bmatrix} + p^3 * \begin{bmatrix} \frac{1}{4} \\ -\frac{1}{4} \end{bmatrix} + p^2 q * \begin{bmatrix} 0 \\ -\frac{1}{2} \end{bmatrix}, \dots$$

Preliminary applications

- Solving systems of linear equations
- Also in the paper
 - Largest eigenvalue
 - PageRank

Contributions

- First steps towards a semantics-preserving notion of fine-grained provenance for linear algebra operators
 - Key development: decomposition, tensor-product construction, and algebraic rules
- Preliminary applications in solving equations, computing largest eigenvalues, and PageRank.
- Key benefit
 - Automatic propagation of annotations through operators
 - Ability to assign values (e.g., 0 or 1) to the annotations and propagate the effects, e.g., for deletion or trust

Related and future work

- Provenance Semirings / Semimodules
 - Green et al. PODS'07, Amsterdamer et al. PODS'11
- Array databases
 - SciDB, RasDaMan
 - Wu et al. SubZero, Peng and Diao SIGMOD'15
- Distributed machine learning / linear algebra programs
 - SystemML, Spark, MLbase, Cumulon, MADlib, GraphX, LINView, etc
- Future work
 - Support more linear algebra operators
 - Scalable implementation

Thank you!