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# Signal Processing on Databases

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Lecture 2: Group Theory

Spreadsheets, Big Tables, and the  
Algebra of Associative Arrays




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# Outline

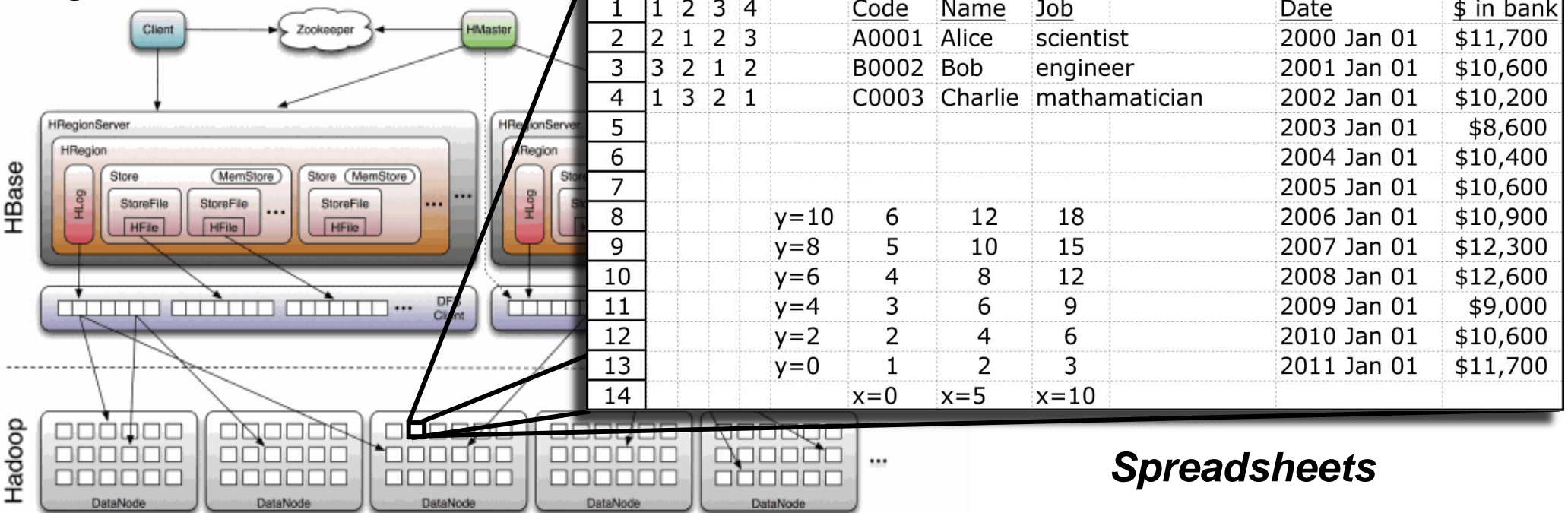
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-  • **Introduction**
  - **What are Spreadsheets?**
  - **Theoretical Goals**
  - **Associative Arrays**
- **Definitions**
- **Group Theory**
- **Vector Space**
- **Linear Algebra**
- **Summary**



# What are Spreadsheets and Big Tables?

## Big Tables



## Spreadsheets

- **Spreadsheets** are the most commonly used analytical structure on Earth (100M users/day?)
- Big Tables (Google, Amazon, Facebook, ...) store most of the analyzed data in the world (Exabytes?)
- **Simultaneous** diverse data: strings, dates, integers, reals, ...
- **Simultaneous** diverse uses: matrices, functions, hash tables, databases, ...
- No formal mathematical basis; Zero papers in AMA or SIAM



# Goal: Signal Processing on Graphs/Strings/Spreadsheets/Tables/ ...

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- **Create a formal basis for working with these data structures based on an Algebra of Associative Arrays**
- **Better Algorithms**
  - Can create algorithms by applying standard mathematical tools (linear algebra and detection theory)
- **Faster Implementation**
  - Associative array software libraries allow these algorithms to be implemented with ~50x less effort
- **Good for managers, too**
  - Much simpler than Microsoft Excel; formally correct



# Multi-Dimensional Associative Arrays

- Extends associative arrays to 2D and mixed data types

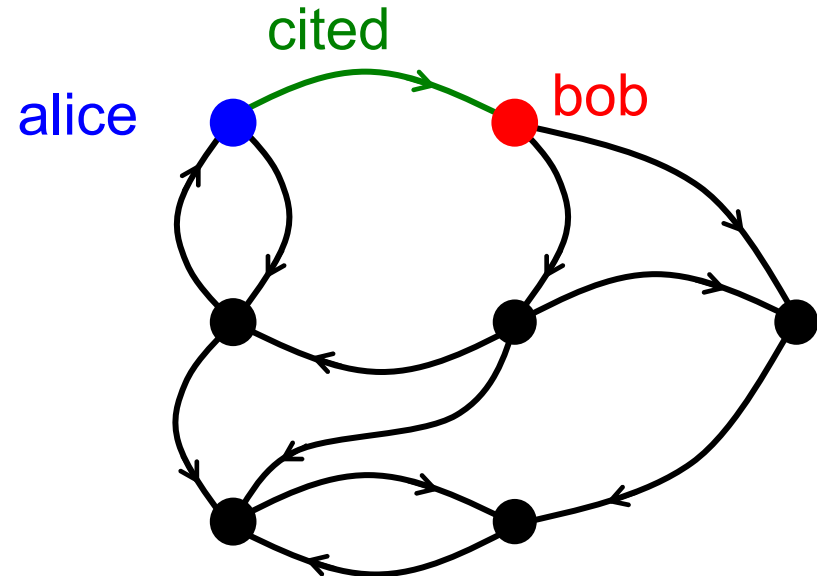
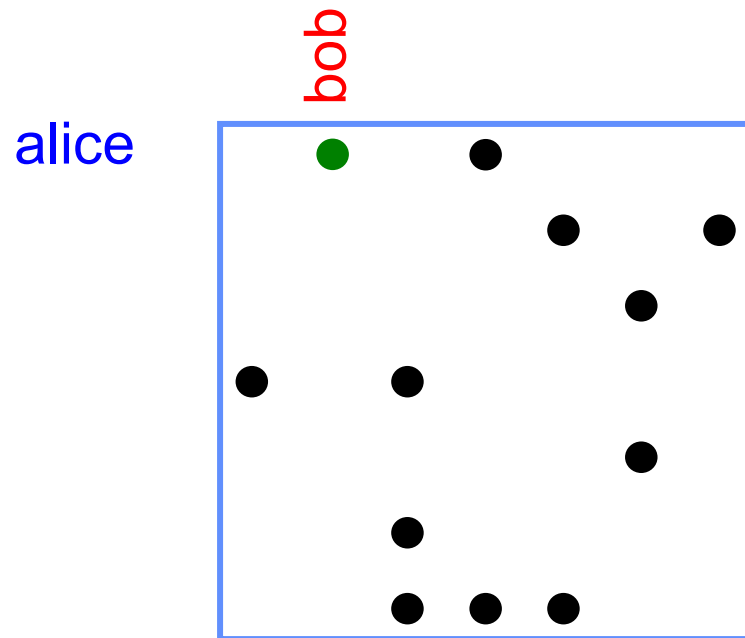
$A('alice ', 'bob ') = 'cited '$

or  $A('alice ', 'bob ') = 47.0$

- Key innovation: 2D is 1-to-1 with triple store

$('alice ', 'bob ', 'cited ')$

or  $('alice ', 'bob ', 47.0)$





# Composable Associative Arrays

- **Key innovation: mathematical closure**
  - All associative array operations return associative arrays

- **Enables composable mathematical operations**

$A + B$     $A - B$     $A \& B$     $A|B$     $A*B$

- **Enables composable query operations via array indexing**

$A('alice\ bob',:)$     $A('alice',:)$     $A('al*',:)$

$A('alice : bob',:)$     $A(1:2,:)$     $A == 47.0$

- **Simple to implement in a library (~2000 lines) in programming environments with: 1<sup>st</sup> class support of 2D arrays, operator overloading, sparse linear algebra**

- **Complex queries with ~50x less effort than Java/SQL**
- **Naturally leads to high performance parallel implementation**



# Universal “Exploded” Schema

Input Data

Time	src_ip	domain	dest_ip
2001-01-01	a		a
2001-01-02	b	b	
2001-01-03		c	c



Triple Store Table: Ttranspose

	2001-01-01	2001-01-02	2001-01-03
src_ip/a	1		
src_ip/b		1	
domain/b		1	
domain/c			1
dest_ip/a	1		
dest_ip/c			1



	src_ip/a	src_ip/b	domain/b	domain/c	dest_ip/a	dest_ip/c
2001-01-01	1				1	
2001-01-02		1	1			
2001-01-03				1		1

Triple Store Table: T

## Key Innovations

- Handles all data into a *single* table representation
- Transpose pairs allows quick look up of *either* row or column



# Outline

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- Introduction
- • Definitions
  - Values
  - Keys
  - Functions
  - Matrix multiply
- Group Theory
- Vector Space
- Linear Algebra
- Summary





# Associative Array Definitions

- Keys and values are from the infinite strict totally ordered set  $\mathbb{S}$
- Associative array  $A(\mathbf{k}) : \mathbb{S}^d \rightarrow \mathbb{S}$ ,  $\mathbf{k}=(k^1, \dots, k^d)$ , is a partial function from  $d$  keys (typically 2) to 1 value, where
$$A(\mathbf{k}_i) = v_i \quad \text{and} \quad \emptyset \text{ otherwise}$$

- Binary operations on associative arrays  $A_3 = A_1 \oplus A_2$ , where  $\oplus = \cup_{f()}$  or  $\cap_{f()}$ , have the properties
  - If  $A_1(\mathbf{k}_i) = v_1$  and  $A_2(\mathbf{k}_i) = v_2$ , then  $A_3(\mathbf{k}_i)$  is
$$v_1 \cup_{f()} v_2 = f(v_1, v_2) \quad \text{or} \quad v_1 \cap_{f()} v_2 = f(v_1, v_2)$$
  - If  $A_1(\mathbf{k}_i) = v$  or  $\emptyset$  and  $A_2(\mathbf{k}_i) = \emptyset$  or  $v$ , then  $A_3(\mathbf{k}_i)$  is
$$v \cup_{f()} \emptyset = v \quad \text{or} \quad v \cap_{f()} \emptyset = \emptyset$$

- High level usage dictated by these definitions
- Deeper algebraic properties set by the collision function  $f()$
- Frequent switching between “algebras” (how spreadsheets are used)



# Associative Array Values

- Value requirements
  - Diverse types: integers, reals, strings, ...
  - Sortable
  - Set
- Let  $\mathbb{S}$  be an infinite strict totally ordered set
  - Total order is an implementation (not theoretical) requirement
  - All values (and keys) will be drawn from this set

- Allowable operations for  $v_1, v_2 \in \mathbb{S}$   
 $v_1 < v_2$                        $v_1 = v_2$                        $v_1 > v_2$

- Special symbols:  $\emptyset, -\infty, +\infty$ 
  - $v \leq +\infty$  is always true ( $+\infty \in \mathbb{S}$ )
  - $v \geq -\infty$  is always true ( $-\infty \in \mathbb{S}$ )
  - $\emptyset$  is the empty set ( $\emptyset \subset \mathbb{S}$ )

**Above properties are consistent with strict totally ordered sets**



# Collision Function $f()$

- Collision function  $f(v_1, v_2)$  can have
  - two contexts ( $\cup \cap$ )
  - three conditions ( $< = >$ )
  - $d + 5$  possible outcomes ( $k v_1 v_2 \emptyset -\infty +\infty$ ) [or sets of these]
- Combinations result in an enormous number of functions ( $\sim 10^{30}$ ) and an even greater number of associative array algebras (function pairs)
  - Impressive level of functionality given minimal assumptions
- Focus on “nice” collision functions
  - Keys are not used inside the function; results are single valued
  - No tests on special symbols

$f(v_1, v_2)$

$v_1 < v_2 : v_1 v_2 \emptyset -\infty +\infty$

$v_1 = v_2 : v \quad \emptyset -\infty +\infty$

$v_1 > v_2 : v_1 v_2 \emptyset -\infty +\infty$

- **Above properties are consistent with strict totally ordered sets**
- **Note:  $\emptyset$  is handled by  $\cup \cap$ ; not passed into  $f()$**



# What About Concatenation?

- Concatenation of values (or keys) can be represented by using  $\cup$  or  $\cap$  as collision function

- Requires generalizing values to sets  $v_1, v_2 \subset \mathbb{S}$

- Allowable operations for  $v_1, v_2 \subset \mathbb{S}$

$$v_1 \cup v_2$$

$$v_1 \cap v_2$$

- Special symbols:  $\emptyset, \mathbb{S}$

$$v \cap \emptyset = \emptyset$$

annihilator (but never reached, so identify)

$$v \cup \mathbb{S} = \mathbb{S}$$

annihilator

$$v \cap \mathbb{S} = v$$

identity

$$v \cup \emptyset = v$$

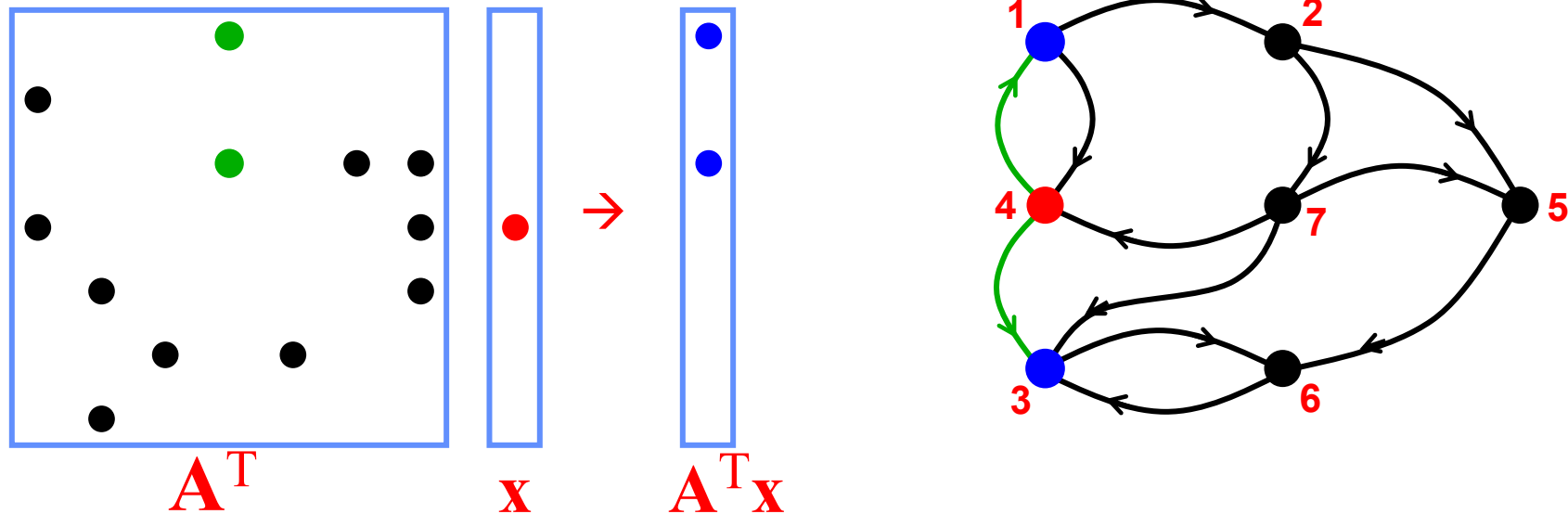
identity

- Possible operators:  $\cup_\cup, \cap_\cup, \cup_\cap, \cap_\cap$

- **Concatenating collision functions are very useful**
- **Can be handled by extending values to be sets**



# Matrix Multiply Framework

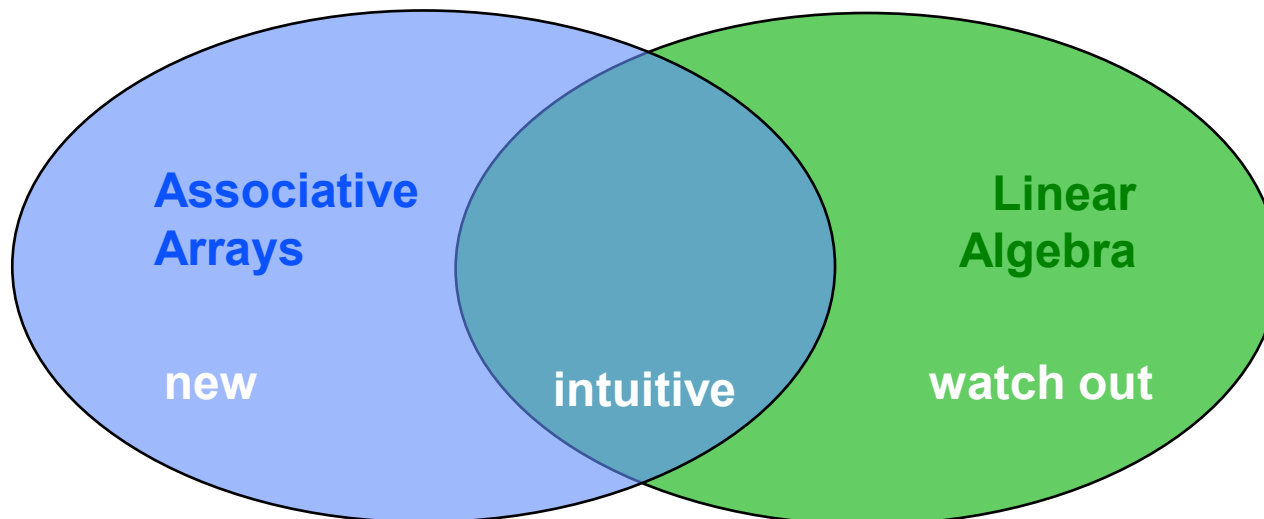


- Graphs can be represented as a sparse matrices
  - Multiply by adjacency matrix  $\rightarrow$  step to neighbor vertices
  - Work-efficient implementation from sparse data structures
- Graph algorithms reduce to products on semi-rings:  $A_3 = A_1 \oplus \cdot \otimes A_2$ 
  - $\otimes$  : associative, distributes over  $\oplus$
  - $\oplus$  : associative, commutative
  - Examples:  $+.*$        $\min.+$        $\text{or.and}$



# Theory Questions

- **Associative arrays can be constructed from a few definitions**
- **Similar to linear algebra, but applicable to a wider range of data**
- **Key questions**
  - **Which linear algebra properties do apply to associative arrays (intuitive)**
  - **Which linear algebra properties do not apply to associative arrays (watch out)**
  - **Which associative array properties do not apply to linear algebra (new)**





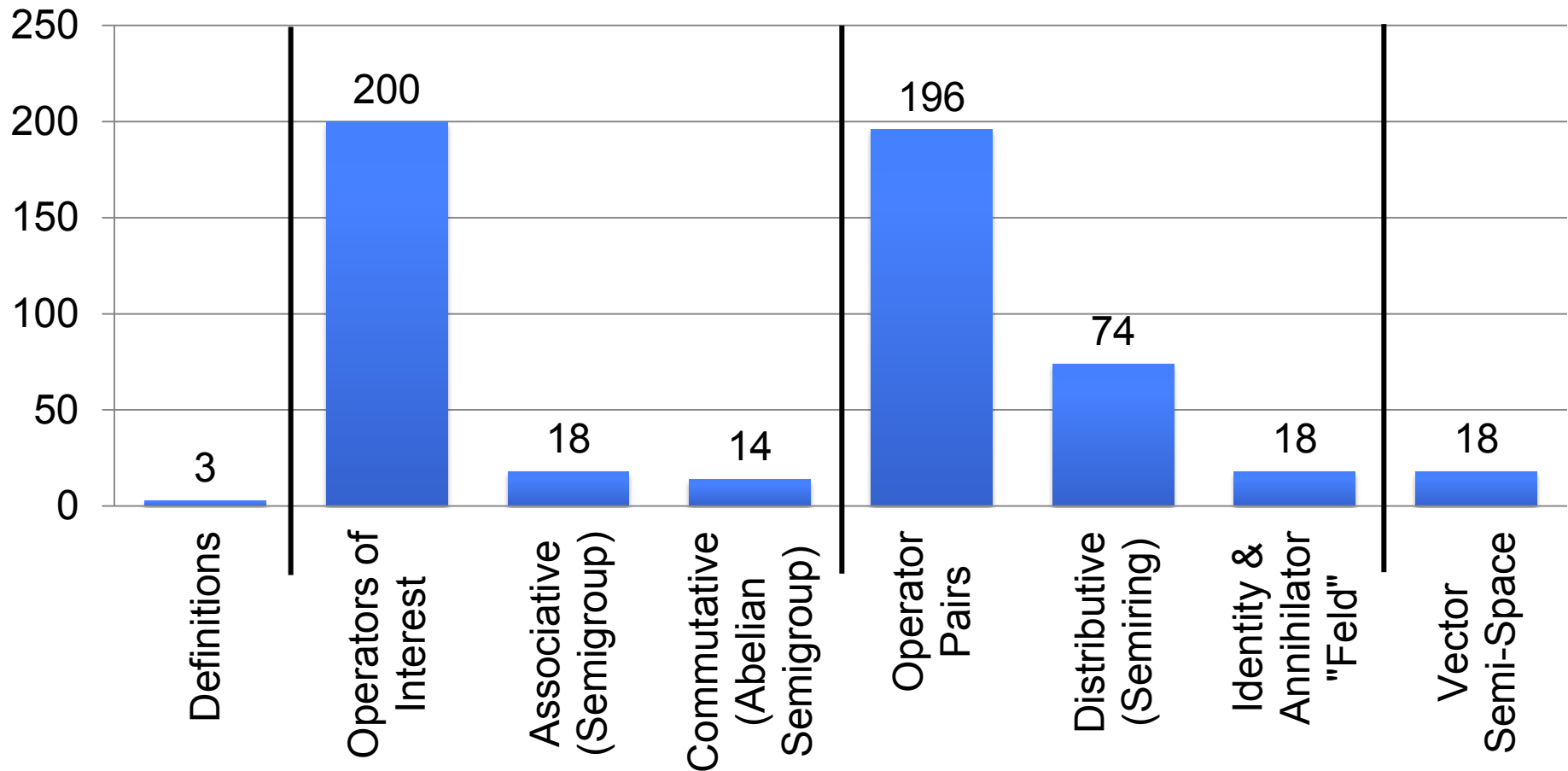
# Outline

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- Introduction
- Definitions
- • **Group Theory**
  - Binary operators
  - Commutative monoids
  - Semirings
  - Feld
- Vector Space
- Linear Algebra
- Summary



# Operators Roadmap

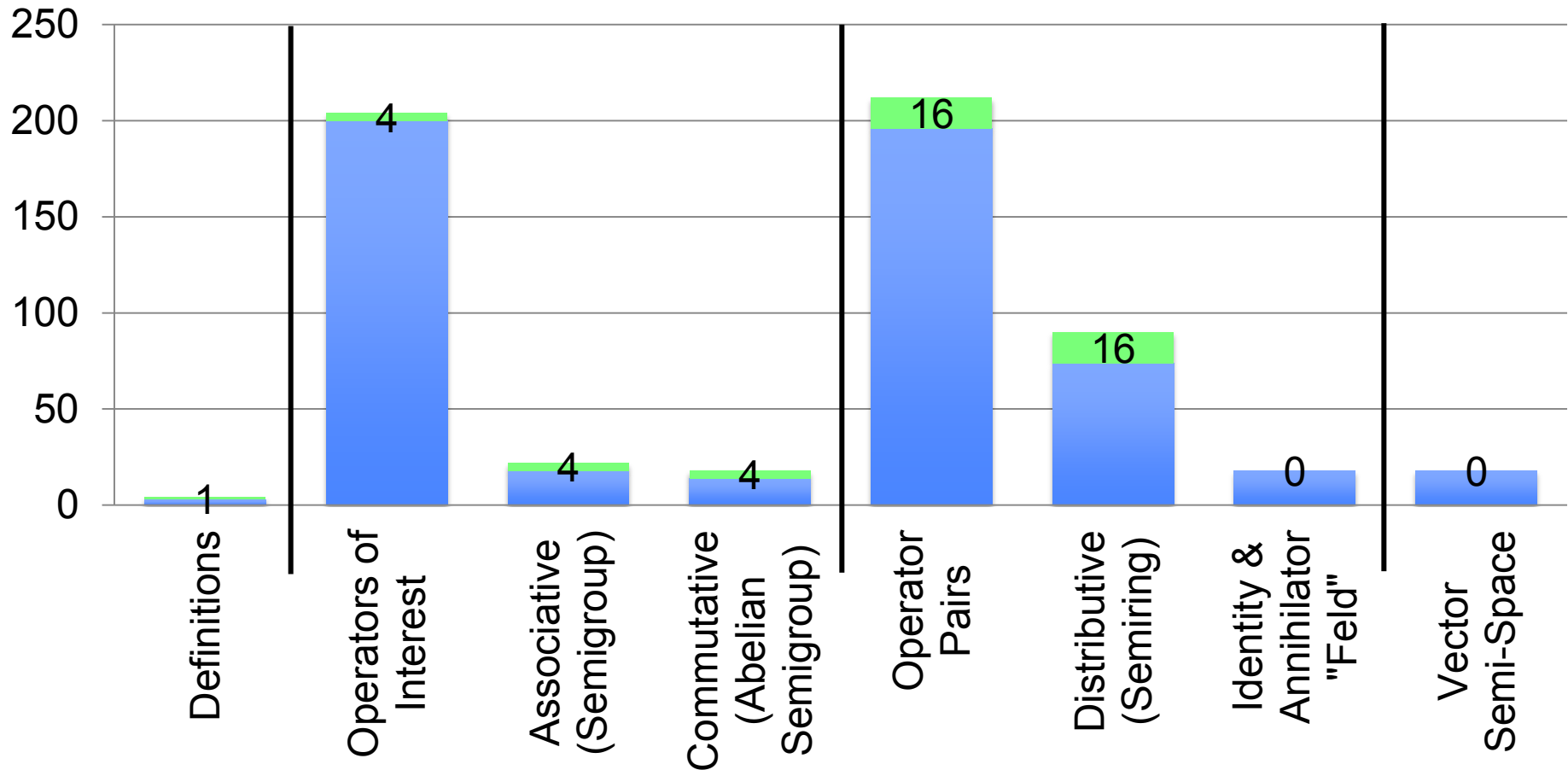


- **Begin with a few definitions**
- **Expand into many operators; reduce to well behaved**
- **Expand into many operator pairs; reduce to well behaved**





# Including Concatenation



- Including concatenation operators expands semirings
- Doesn't expand vector semi-space



# Associative and Commutative Operators

ID	Operator $\oplus$	$v_1 < v_2$	$v_1 = v_2$	$v_1 > v_2$
1	$\cup_{\text{left}}$	$v_1$	$v$	$v_1$
2	$\cap_{\text{left}}$	$v_1$	$v$	$v_1$
3	$\cup_{\text{max}}$	$v_2$	$v$	$v_1$
4	$\cap_{\text{max}}$	$v_2$	$v$	$v_1$
41	$\cup_{\text{min}}$	$v_1$	$v$	$v_2$
42	$\cap_{\text{min}}$	$v_1$	$v$	$v_2$
43	$\cup_{\text{right}}$	$v_2$	$v$	$v_2$
44	$\cap_{\text{right}}$	$v_2$	$v$	$v_2$
86	$\cap_{\delta}$	$\emptyset$	$v$	$\emptyset$
96	$\cap_{\emptyset}$	$\emptyset$	$\emptyset$	$\emptyset$
127	$\cup_{-\infty, \delta}$	$-\infty$	$v$	$-\infty$
128	$\cap_{-\infty, \delta}$	$-\infty$	$v$	$-\infty$
147	$\cup_{-\infty}$	$-\infty$	$-\infty$	$-\infty$
148	$\cap_{-\infty}$	$-\infty$	$-\infty$	$-\infty$
169	$\cup_{+\infty, \delta}$	$+\infty$	$v$	$+\infty$
170	$\cap_{+\infty, \delta}$	$+\infty$	$v$	$+\infty$
199	$\cup_{+\infty}$	$+\infty$	$+\infty$	$+\infty$
200	$\cap_{+\infty}$	$+\infty$	$+\infty$	$+\infty$

- Associative

$$(v_1 \oplus v_2) \oplus v_3 = v_1 \oplus (v_2 \oplus v_3)$$

- 18 associative operators
  - Semigroups
  - Groups w/o inverses

- Commutative

$$v_1 \oplus v_2 = v_2 \oplus v_1$$

- 14 associative & commutative operators
  - Removes left and right
  - Abelian Semigroups
  - Abelian Groups w/o inverses



# Distributive Operator Pairs

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- 14 x 14 = 196 Pairs of Abelian Semigroup operators

- Distributive

$$v_1 \otimes (v_2 \oplus v_3) = (v_1 \otimes v_2) \oplus (v_1 \otimes v_3)$$

- 74 distributive operator pairs
  - Semirings
  - Rings without inverses and without identity elements

- **1/3 of possible operator pairs are semirings**



# Distributive Operator Pairs with Annihilators (0) and Identities (1)

- $\oplus$  identity:  $v_1 \oplus 0 = v_1$        $0 = \emptyset, -\infty, +\infty$
- $\otimes$  identity:  $v_1 \otimes 1 = v_1$        $1 = \emptyset, -\infty, +\infty$
- $\otimes$  annihilator:  $v_1 \otimes 0 = 0$        $0 = \emptyset, -\infty, +\infty$
  
- 12 Semirings with appropriate 0 1 set (4 with two)
- 16 total over six operators:  $\cup_{\max}, \cap_{\max}, \cup_{\min}, \cap_{\min}, \cup_{-\infty}, \cup_{+\infty}$ 
  - Fields? (Fields w/o inverses)
  
- $\oplus = \cup_{f()}$  in 10/16 ( $\cup$  feels more like plus)
- $\otimes = \cap_{f()}$  in 10/16 ( $\cap$  feels more like multiply)
- $\oplus = \cup_{f()}$  and  $\otimes = \cap_{f()}$  in 8/16
- $0 = \emptyset$  in 6/8 ( $\emptyset$  feels more like zero,  $0 > 1$  might be a problem)

• **1/5 of semirings are Fields (Fields w/o inverses)**



# Operator Pairs



	0	1	$\cup_{\max}$	$\cap_{\max}$	$\cup_{\min}$	$\cap_{\min}$	$\cap_{\delta}$	$\cap_{\emptyset}$	$\cup_{-\infty, \delta}$	$\cap_{-\infty, \delta}$	$\cup_{-\infty}$	$\cap_{-\infty}$	$\cup_{+\infty, \delta}$	$\cap_{+\infty, \delta}$	$\cup_{+\infty}$	$\cap_{+\infty}$
$\cup_{\max}$			D	$\emptyset -\infty$		$-\infty +\infty$ $\emptyset +\infty$		D				D			D	D
$\cap_{\max}$				D	$-\infty +\infty$ $-\infty \emptyset$	$-\infty +\infty$		D			$-\infty \emptyset$	D				D
$\cup_{\min}$			$+\infty -\infty$ $+\infty \emptyset$	$+\infty -\infty$ $\emptyset -\infty$	D	$\emptyset +\infty$		D			D	D				D
$\cap_{\min}$				$+\infty -\infty$		D		D				D			$+\infty \emptyset$	D
$\cap_{\delta}$							D	D								
$\cap_{\emptyset}$				D		D	D	D		D		D		D		D
$\cup_{-\infty, \delta}$								D	D	D	D	D				D
$\cap_{-\infty, \delta}$								D		D		D				D
$\cup_{-\infty}$						$\emptyset +\infty$		D		D		D				
$\cap_{-\infty}$						D		D		D		D				
$\cup_{+\infty, \delta}$								D				D	D	D	D	D
$\cap_{+\infty, \delta}$								D				D		D		D
$\cup_{+\infty}$				$\emptyset -\infty$				D						D		D
$\cap_{+\infty}$				D				D						D		D

D=distributes; 0=Plus Identity/Multiply Annihilator; 1=Multiply Identity



# Concatenate Operators

ID	Operator $\oplus$	$f(v_1, v_2)$
201	$\cup_{\cup}$	$v_1 \cup v_2$
202	$\cap_{\cup}$	$v_1 \cup v_2$
203	$\cup_{\cap}$	$v_1 \cap v_2$
204	$\cap_{\cap}$	$v_1 \cap v_2$

- Recall  $v_1$  and  $v_2$  are sets
- All operators are associative and commutative
  - 4 Abelian Semigroups

0 1	$\cup_{\cup}$	$\cap_{\cup}$	$\cup_{\cap}$	$\cap_{\cap}$
$\cup_{\cup}$	D	$\emptyset -\infty$	D	$-\infty +\infty$ $\emptyset +\infty$
$\cap_{\cup}$	D	D	$-\infty +\infty$ $-\infty \emptyset$	$-\infty +\infty$
$\cup_{\cap}$	$+\infty -\infty$ $+\infty \emptyset$	$+\infty -\infty$ $\emptyset -\infty$	D	$\emptyset +\infty$
$\cap_{\cap}$	D	$+\infty -\infty$	D	D

- All operator pairs distribute
  - 16 Semirings



# Outline

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- **Introduction**
- **Definitions**
- **Group Theory**
- • **Vector Space**
  - **Vector Semispace**
  - **Uniqueness**
- **Linear Algebra**
- **Summary**



# Vector Space over a Feld

- Associative Array Vector  $\oplus$ 
    - All associative arrays are conformant (unlike matrices)
  - Associative Array Scalar  $\otimes$ 
    - Scalar is a value applied directly to values; similar to constant function; or a function that takes on keys of non-scalar argument
  - Vector Space  $\oplus$  requirements
    - Commutes [Yes]; Associative [Yes]; 0 Identity element [Yes]
    - Inverse [No]
  - Vector Space scalar  $\otimes$  requirements
    - Commutes [Yes]; Associative [Yes]; Distributes over addition [Yes]; 1 Identity element [Yes]
- All associative array operator pairs that yield Felds also result in Vector Spaces wo/inverses (Vector Semispace?)**





# Vector Semispace Properties

- Scalar  $\oplus$  identity annihilates under  $\otimes$  [Yes]
- Subspace [Yes]
  - Any linear combination of vectors taken from the subspace is in the subspace and obeys the properties of a vector space
  - Theorem: Intersection of any subspaces is a subspace?
- Span [Yes+]
  - Given a set of vectors  $A_j$ , their span is all linear combinations of those vectors (includes vectors of different lengths)
$$\oplus_j (a_j \otimes A_j)$$
- Span = Subspace [Yes?]
  - Given an arbitrary set of vectors, their span is a vector space?
- Linear dependence [No]
  - There is a non-trivial linear combination of vectors equal to the  $\oplus$  identity; can't do this without additive inverse
  - Need to redefine linear independence or all vectors are linearly independent; use minimum vectors in a subspace definition?
  - Likewise need to redefine basis as it depends upon linear dependence

**Key question: under what conditions does the result of a linear combination of associative arrays uniquely determine the coefficients**



# Unique Coefficient Conditions

- Consider a linear combinations of two associative array vectors

$$A_3 = (a_1 \otimes A_1) \oplus (a_2 \otimes A_2)$$

- Let  $\oplus = \cup_{\min}$ ,  $\otimes = \cap_{\max}$ ,  $0 = \emptyset$ , and  $1 = -\infty$
- When are  $a_1$  and  $a_2$  uniquely determined by  $A_1$ ,  $A_2$  and  $A_3$  ?

<u>Canonical Vectors</u>	<u>Single valued</u>	<u>Multi-valued</u>
$A_1(k_1) = -\infty$ $A_2(k_2) = -\infty$		$A_1(k_1 k_2) = (v_1 v_2)$ $A_2 = A_1$ $v_1 < v_2$
$A_1(k_1) = +\infty$ $A_2(k_2) = +\infty$	$A_1(k_1 k_2) = (v v)$ $A_2 = A_1$	$A_1(k_1 k_2) = (v_1 v_2)$ $A_2(k_1 k_2) = (v_2 v_1)$ $v_1 < v_2$

- Consider specific cases to show existence of uniqueness**



# Canonical Vectors

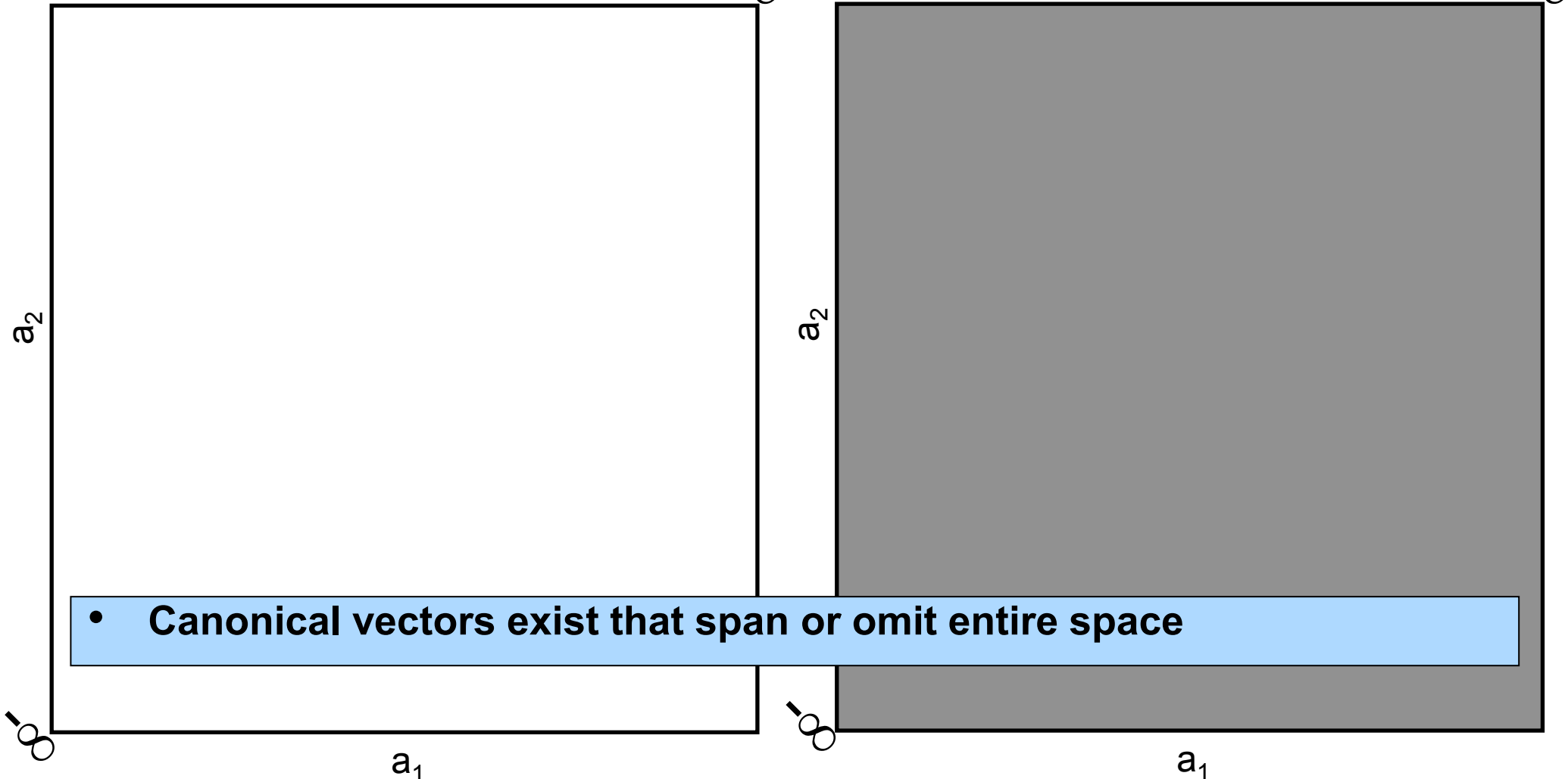
$A_1(k_1) = -\infty$

$A_2(k_2) = -\infty$


 $x \in \mathbb{R}^2$ 


$A_1(k_1) = +\infty$


$A_2(k_2) = +\infty$


 $x \in \mathbb{R}^2$ 

• Canonical vectors exist that span or omit entire space

  $a_1, a_2$  unique

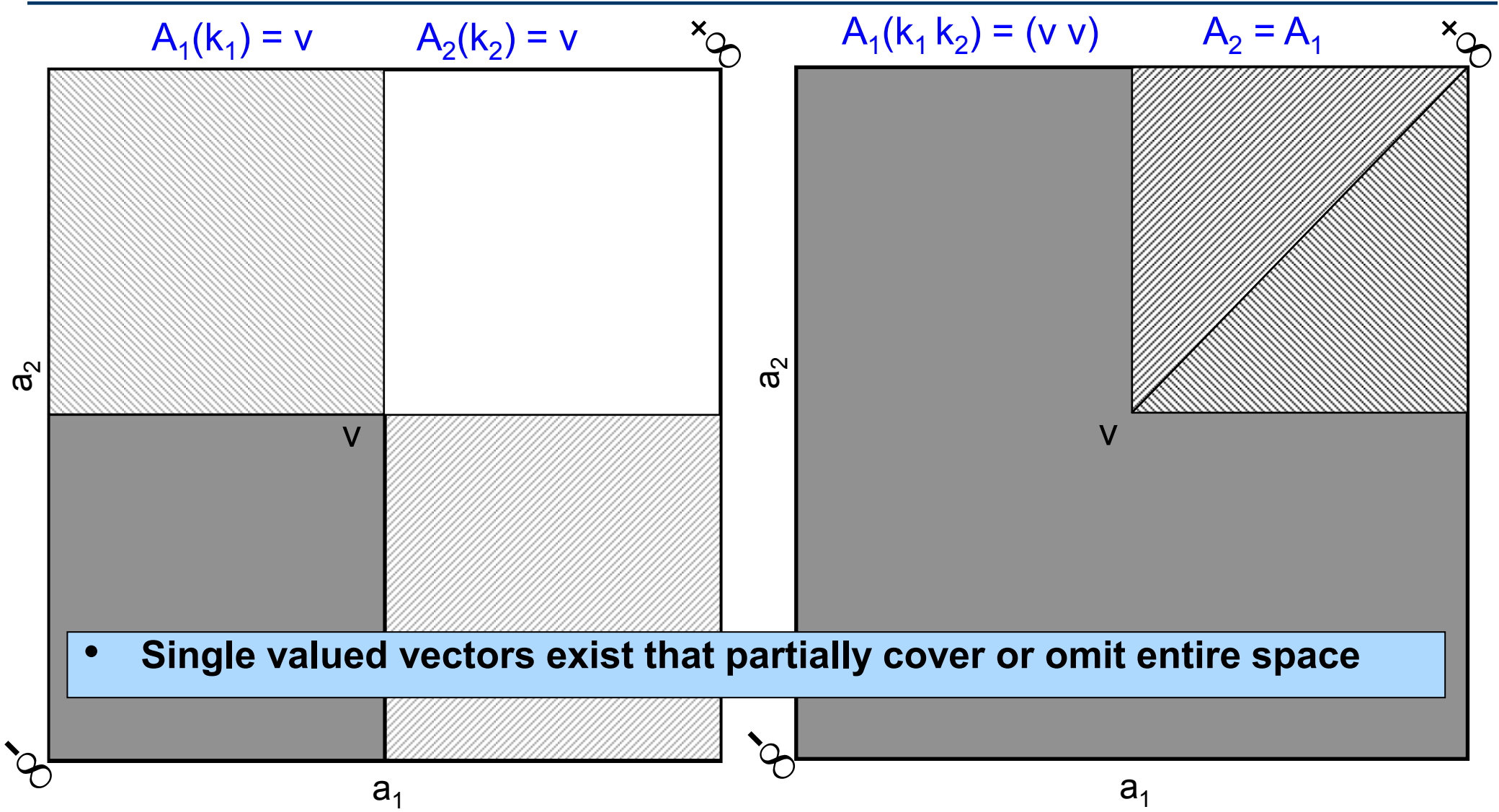
  $a_1$  unique

  $a_2$  unique

  $a_1, a_2$  not unique



# Single Valued Vectors



• Single valued vectors exist that partially cover or omit entire space

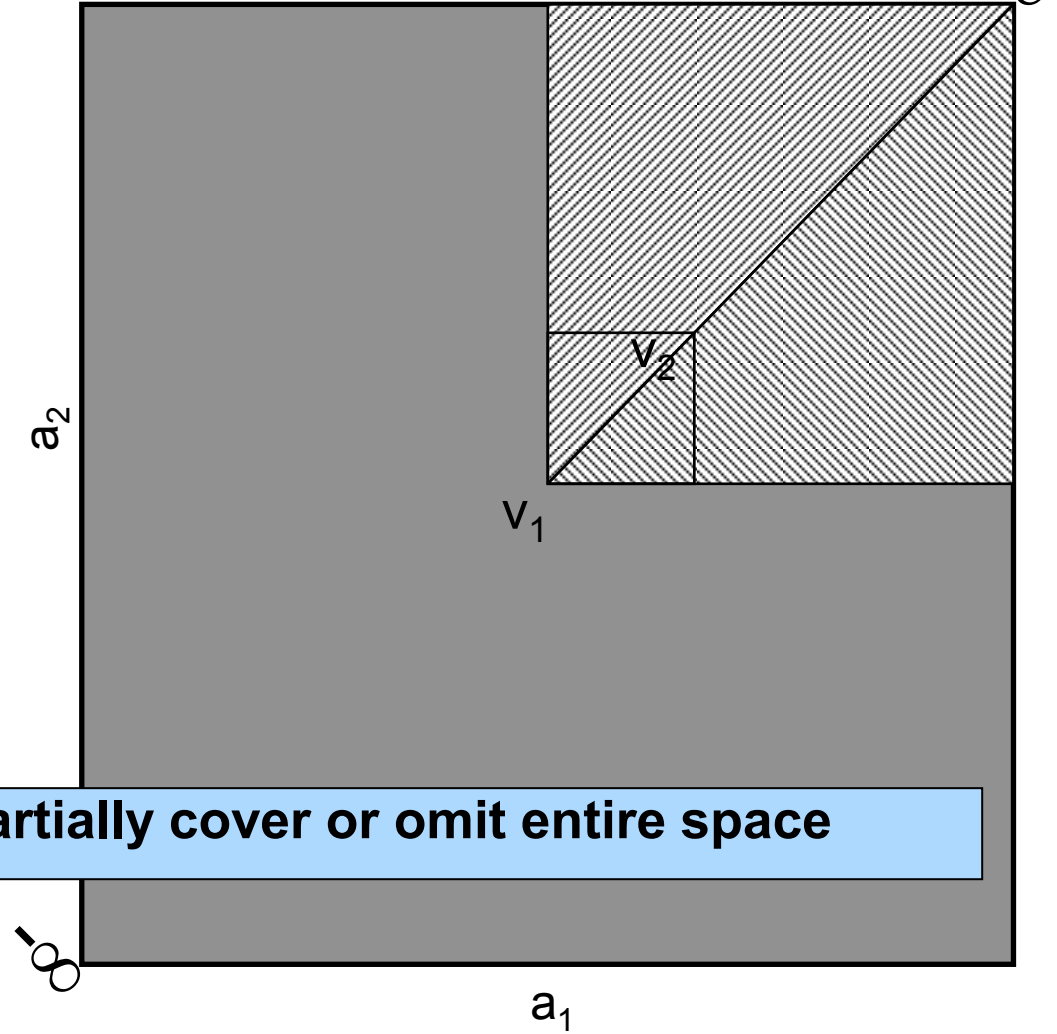
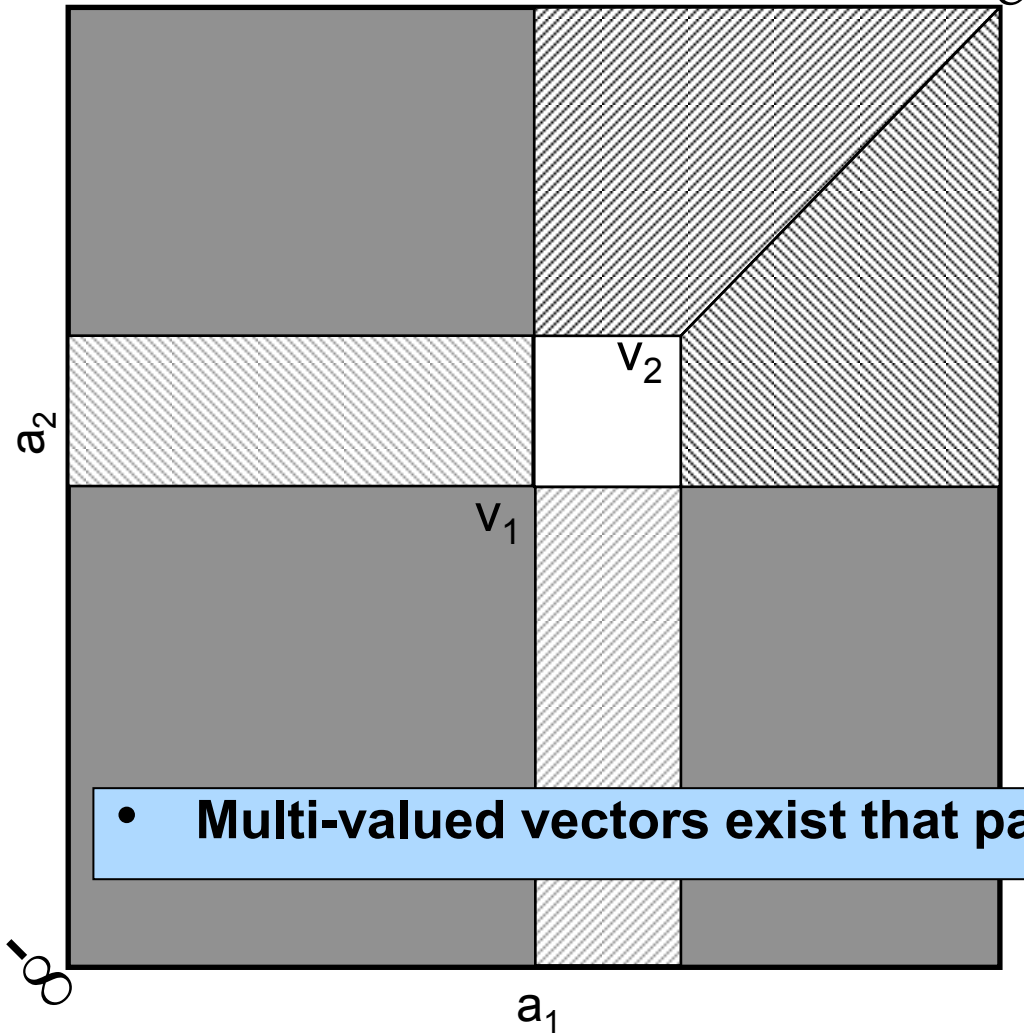
- $a_1, a_2$  unique
- $a_1$  unique
- $a_2$  unique
- $a_1, a_2$  not unique



# Multi-Valued Vectors

$$A_1(k_1, k_2) = (v_1, v_2), A_1(k_1, k_2) = (v_2, v_1), v_1 < v_2$$

$$A_1(k_1, k_2) = (v_1, v_2), A_2 = A_1, v_1 < v_2$$



• Multi-valued vectors exist that partially cover or omit entire space

$a_1, a_2$  unique

$a_1$  unique

$a_2$  unique

$a_1, a_2$  not unique



# Outline

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- **Introduction**
- **Definitions**
- **Group Theory**
- **Vector Space**
- • **Linear Algebra**
  - **Transpose**
  - **Special Matrices**
  - **Matrix Multiply**
  - **Identity**
  - **Inverses**
  - **Eigenvectors**
- **Summary**



# Matrix Transpose

- Swap keys (rows and columns)

$$A(r,c)^T = A(c,r)$$

- No change with even number of transposes
- Transpose distributes across  $\oplus$  and scalar  $\otimes$

$$((a_1 \otimes A_1) \oplus (a_2 \otimes A_1))^T = (a_1 \otimes A_1^T) \oplus (a_2 \otimes A_1^T)$$

- **Similar to linear algebra**



# Special Matrices

- Submatrices [Yes]
  - Zero matrix [Yes?] (empty set)
  - Square matrix [Yes]
  - Diagonal matrix [Yes]
  - Upper/lower triangular [Yes]
  - Skew symmetric [No] (no  $\oplus$  inverse)
  - Hermitian [No] (no  $\oplus$  inverse)
  - Elementary row/column operations [Yes?]
    - Swap both keys or values? No  $\otimes$  inverse.
    - If both key and value swap, then equivalent to matrix multiply
  - Row/column equivalence [Yes?]
    - If limit to swaps
- 
- **Similar and different from linear algebra**
  - **Possible to construct these forms, but may not be applicable to associative arrays that have fixed keys (i.e., functions over a keys)**





# Matrix Multiply

- Matrix multiply

$$A_3 = A_1 A_2 = A_1 \oplus \cdot \otimes A_2$$

- Always conformant (can multiply any sizes)
- Inner product formulation (computation)

$$A_3(r_i, c_j) = \oplus_k ( A_1(r_i, k) \otimes A_2(k, c_j) )$$

- Outer product formulation (theory)

$$A_k(r_i, c_j) = A_1(r_i, k) \otimes A_2(k, c_j)$$

$$A_3 = \oplus_k A_k$$

- **Different from linear algebra**
- **Associative arrays have no conformance requirements**



# Matrix Multiply Examples

- 1x2 Row matrix:  $A_1(r, k_1 k_2) = v_1$
- 2x1 Column matrix:  $A_1(k_2 k_3, c) = v_2$
- Example 1: 1x1 Matrix:  $A_3(r, c) = A_1 A_2 =$  [See Table]
- Example 2: 2x2 Matrix ( $r \neq c$ ):  $A_3(k_1 k_2, k_2 k_3) = A_2 A_1 =$  [See Table]
- Example 3: 2x2 Matrix ( $r = c$ ):  $A_3(k_1 k_2, k_2 k_3) = A_2 A_1 = f(v_1, v_2)$
- Value of  $A_3$  depends upon specifics of  $\oplus$  and  $\otimes$

Example 1	$\otimes = \cup_{f()}$	$\otimes = \cap_{f()}$
$\oplus = \cup_{g()}$	$g(g(v_1, f(v_1, v_2), v_2))$	$f(v_1, v_2)$
$\oplus = \cap_{g()}$	$g(g(v_1, f(v_1, v_2), v_2))$	$\emptyset$

Example 2	$\otimes = \cup_{f()}$	$\otimes = \cap_{f()}$
$\oplus = \cup_{g()}$	$g(v_1, v_2)$	$\emptyset$
$\oplus = \cap_{g()}$	$g(v_1, v_2)$	$\emptyset$

**Wide range of behaviors possible given specific operator choices**



# Identity

- Left Identity:  $I_{\text{left}} = \text{diag}(\text{Row}(A)) = 1$
- When does?  $I_{\text{left}} A = A$
- Right Identity:  $I_{\text{right}} = \text{diag}(\text{Col}(A)) = 1$
- When does?  $A I_{\text{right}} = A$

- Generally possible when

$$\oplus = \cup_{g()}\quad \otimes = \cap_{f()}$$

- In some circumstances

$$I = I_{\text{left}} \oplus I_{\text{right}} \quad \text{and} \quad A I = A = I A$$

- **Similar to linear algebra for a limited set of  $\oplus$  and  $\otimes$**



# Inverses

- Left Inverse:  $A A^{-1} = I_{\text{left}}$
- Right Inverse:  $A^{-1} A = I_{\text{right}}$
- Is it possible to construct matrix inverses with no  $\oplus$  inverse and no  $\otimes$  inverse
- Generally, no. Exception
  - $A$  is a column/row vector
  - $\oplus = \cup_{g()}$ ,  $\otimes = \cap_{f()}$
  - $I_{\text{right/left}}$  is 1x1 equal to “local” 1 (i.e., 1 wrt to  $A$ )

- **Different from linear algebra**
- **Inverses generally do not appear in associative arrays**



# Eigenvectors (simple case)

- Let  $\oplus = \cup_g$ ,  $\otimes = \cap_f$
- Let  $A$ ,  $A_e$ ,  $A_\lambda$  be  $N \times N$  and have 1 element per row and column

$$A(r_i, r_i) = v_i$$

$$A_e(r_i, c_i) = e_i$$

$$A_\lambda(c_i, c_i) = v_i$$

- Eigenvector equation

$$A A_e = A_e A_\lambda = A_{e\lambda}$$

- where:  $A_{e\lambda}(r_i, c_i) = f(v_i, e_i)$

- **Eigenvector equation satisfied in a simple case**
- **Row and column keys must match**



# Pseudoinverse (simple case)

- Let  $\oplus = \cup_g$ ,  $\otimes = \cap_f$
- Let  $A$ ,  $A^+$  be  $N \times N$  (or  $N_r \times N_c$ ?) and have 1 element per row and column

$$A(r_i, c_i) = v_i$$

$$A^+(c_i, r_i) = v_i^+$$

- Pseudoinverse requires

$$A = A A^+ A$$

$$A = A^+ A A^+$$

$$(A A^+)^T = A A^+$$

$$(A A^+)^T = A A^+$$

- where:  $f(v_i, v_i^+) = v_i$

- Pseudoinverse equation satisfied in a simple case
- Row and column keys can be different



# Future Work: Got Theorems?

- **Spanning theorems: when is a span a vector space?**
  - **Linear dependence: adding a vector doesn't change span?**
  - **Identity Array: when do left/right identity exist?**
  - **Inverse: why doesn't it exist?**
  - **Determinant: existence?**
  - **Pseudoinverse: existence? How to compute?**
  - **Linear transforms: existence?**
  - **Norms or inner product space**
  - **Compressive sensing requirements**
  - **Eigenvectors**
  - **Convolution (with next operator)**
  - **Complementary matrices**
- **For which  $\oplus$ ,  $\otimes$ , 0/1 do these apply**



# Summary

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- **Algebra of Associative Arrays provides the mathematics for representing and operating on Spreadsheets and Big Tables**
- **Small number of assumptions yields a rich mathematical environment**
- **Much of linear algebra is available without  $\oplus$  inverse and  $\otimes$  inverse**





# Example Code & Assignment

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- **Example Code**
  - **d4m\_api/examples/1Intro/3GroupTheory**
  
- **Assignment 2**
  - **Define, in words, a list of operations that make “sense” for your associative arrays in Assignment 1**
  - **Explain your reasoning**



# Relational Model High Level Comparison

	Relational Database	Associative Arrays
Fill	Dense	Sparse
Columns	Static	Dynamic
Data	Typed	Untyped
#Rows	Unlimited	Unlimited
#Columns	Small	Unlimited
Dimensions	2 different	N same
Main Operation	Join	Linear Algebra

- Relational algebra (Codd 1970) is the de facto theory of databases
- The design goal of relational algebra and associative arrays algebra are fundamentally different
- Result in a fundamental differences in the theory

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