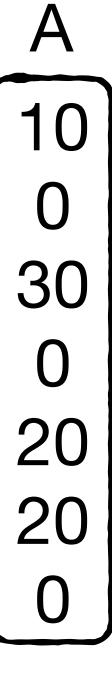
# Column-Store Tutorial

CSC443H1 Database System Technology

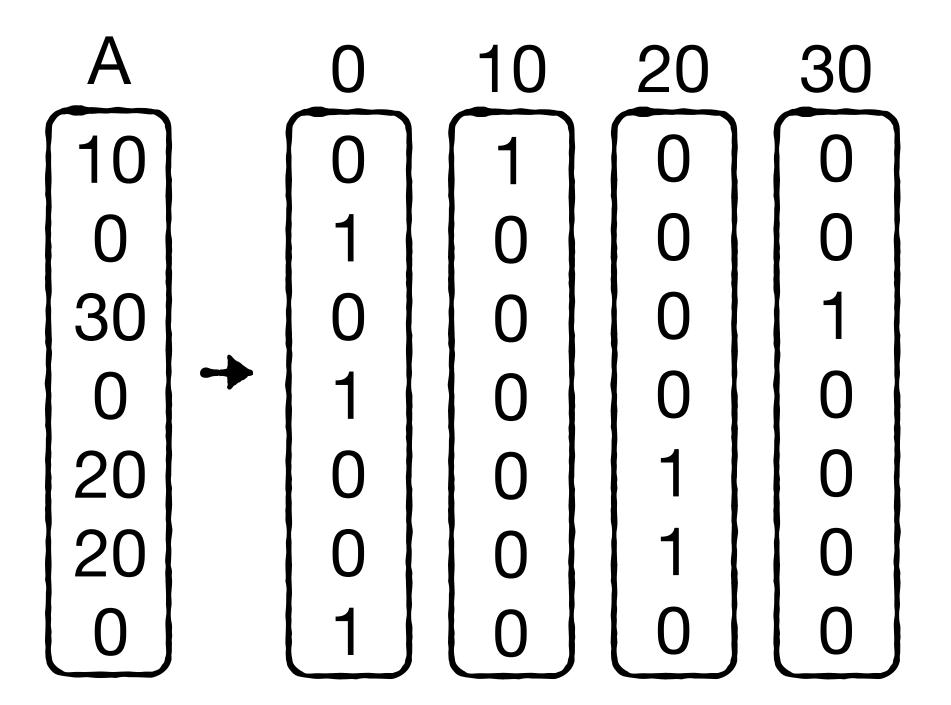
Column A has N entries of which |A| are unique and |A| << N. It is subject to: "select B where  $x \le A$  and  $y \ge A$ " where x < y. We want to compress the column and process the above query without decompressing the data.

- (1) How do we achieve this when  $1 < |A| < \log_2(N)$ ?
- (2) How about when  $\log_2(N) < |A| < \sqrt{N}$ ?



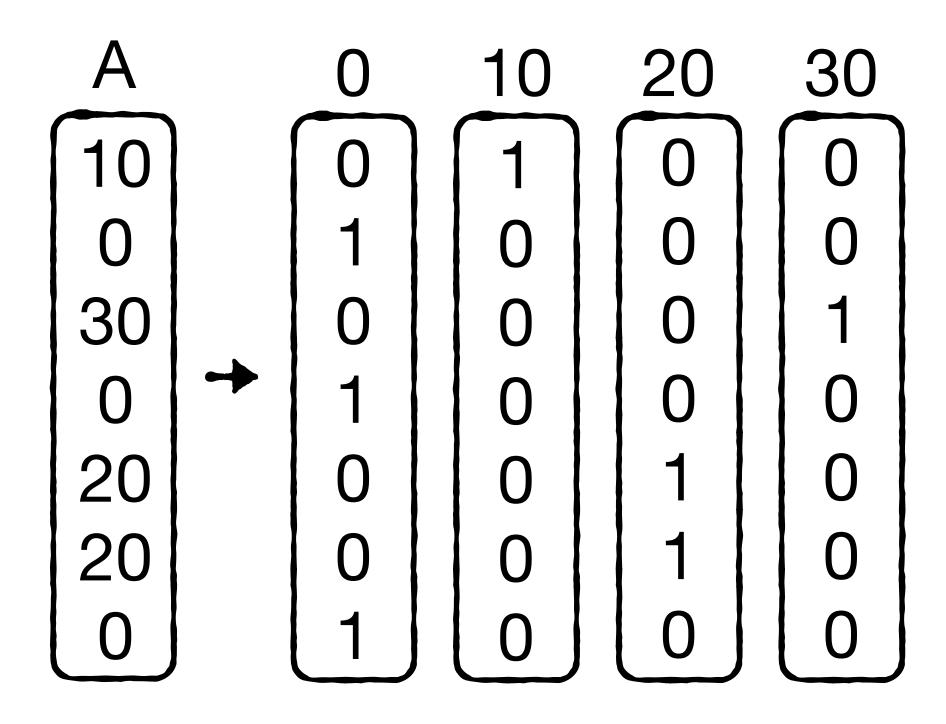
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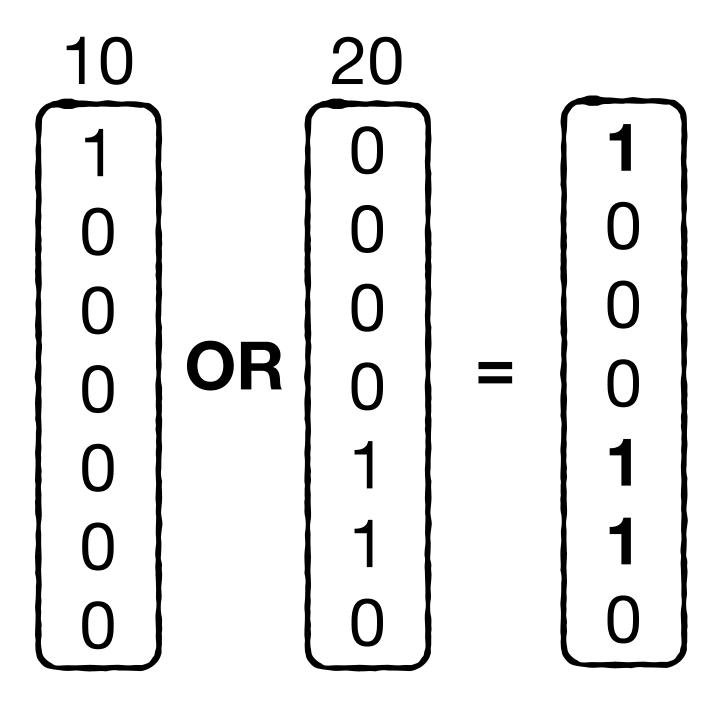


#### How to process queries?

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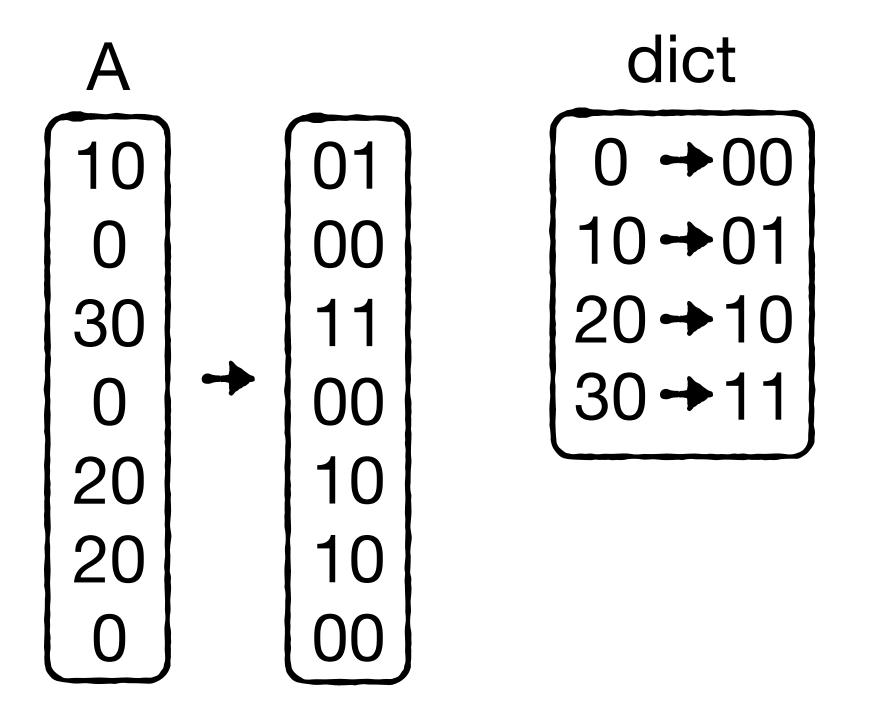
e.g., select B where  $10 \le A$  and  $20 \ge A$ 

"OR" the vectors of values in the range.

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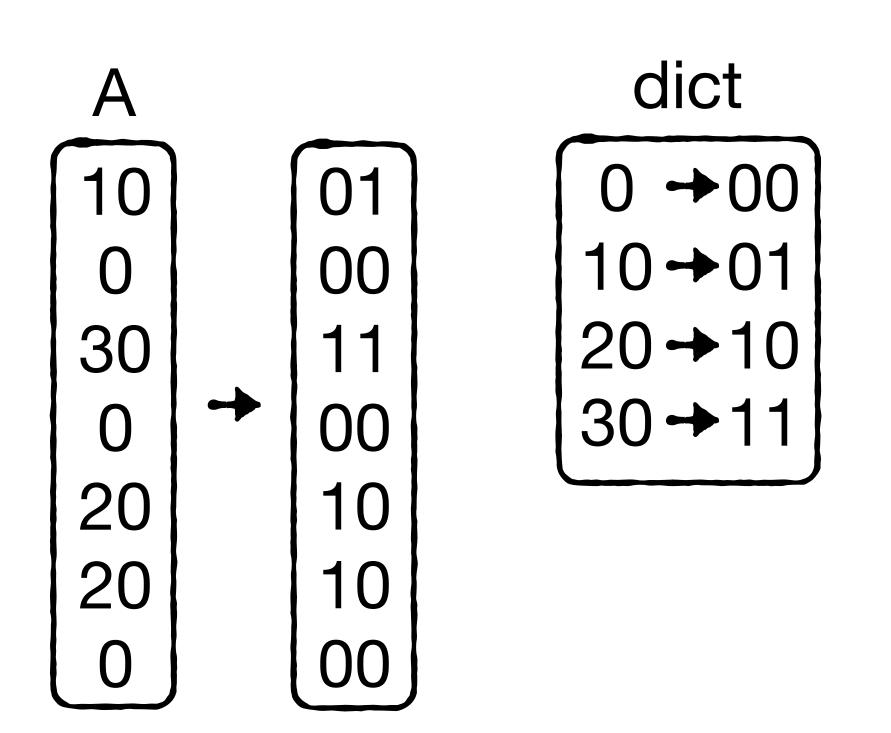
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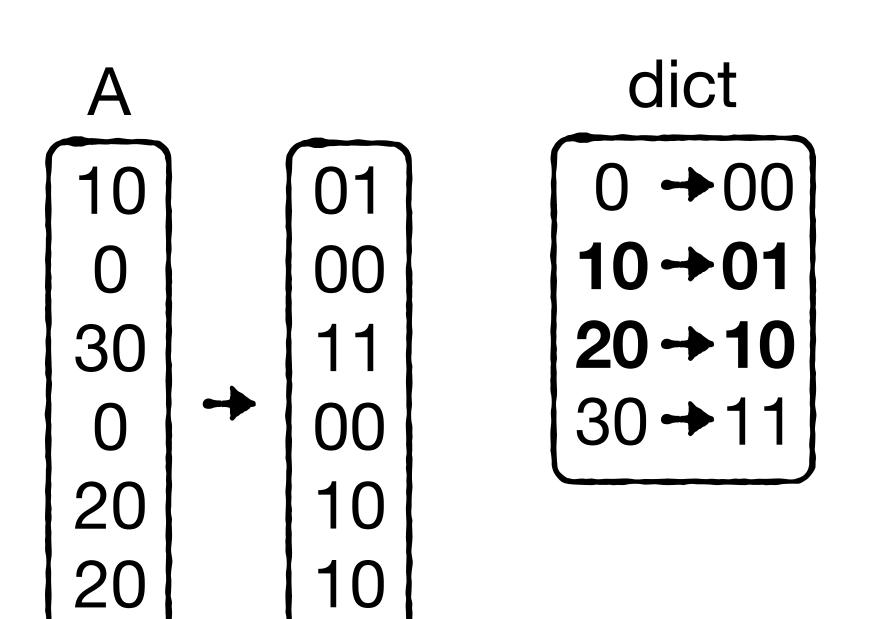
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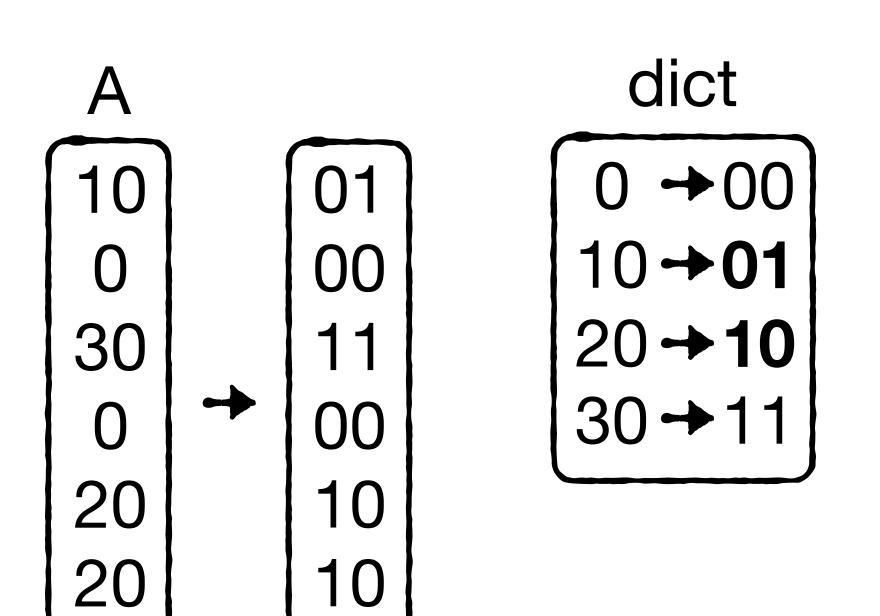
e.g., select B where 10 ≤ A and 20 ≥ A

↑

Replace dictionary values in query

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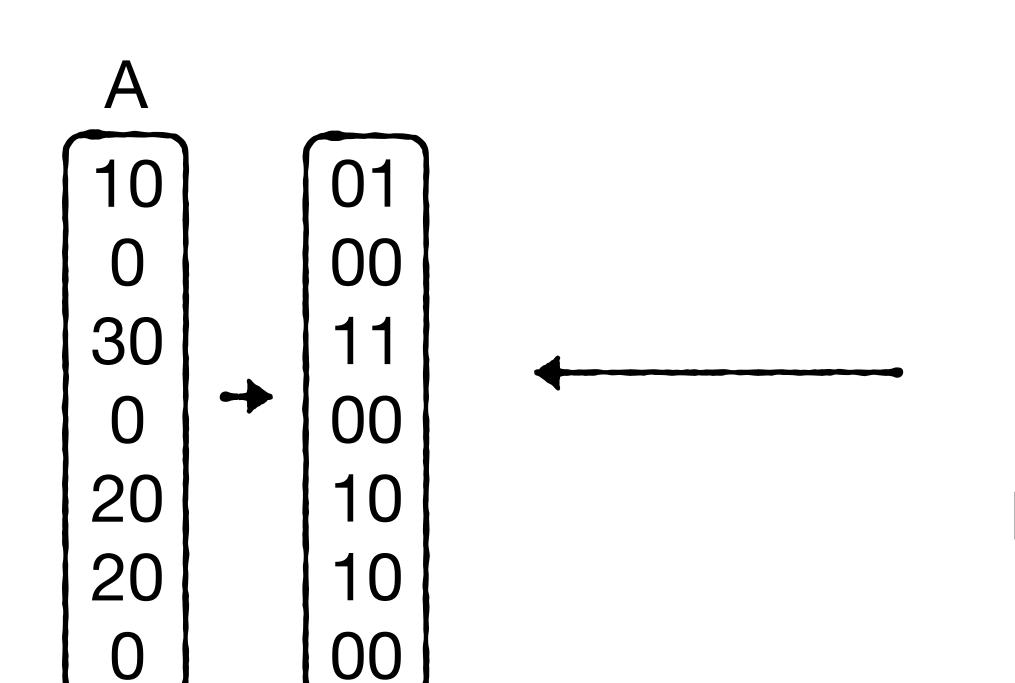
e.g., select B where 01 ≤ A and 10 ≥ A

↑

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How to process queries?

e.g., select B where 01 ≤ A and 10 ≥ A

Run on compressed column.

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What happens if  $|A| > \sqrt{N}$  ?

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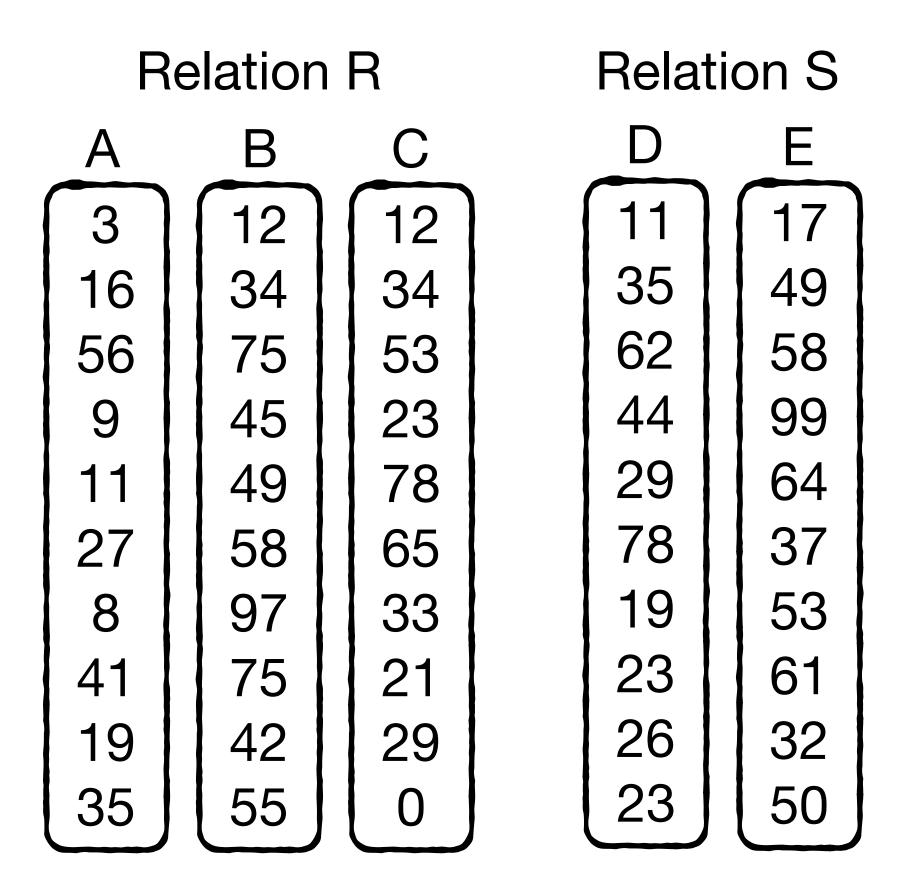
What happens if  $|A| > \sqrt{N}$  ?

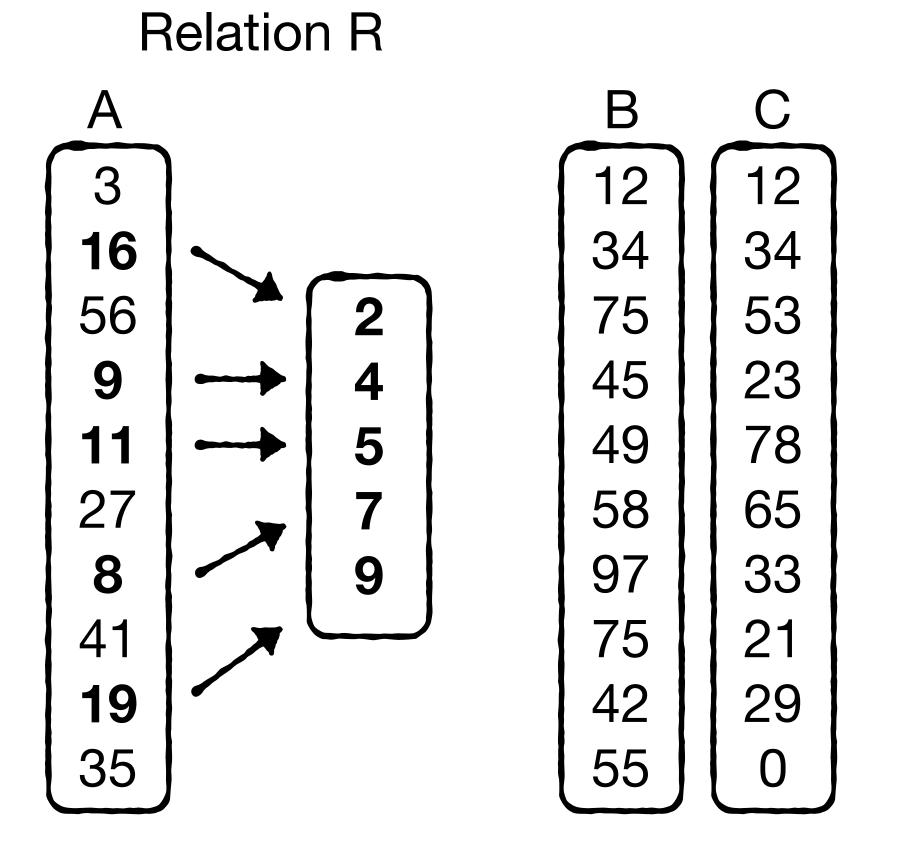
Size of each code becomes at least: log<sub>2</sub>(N) / 2

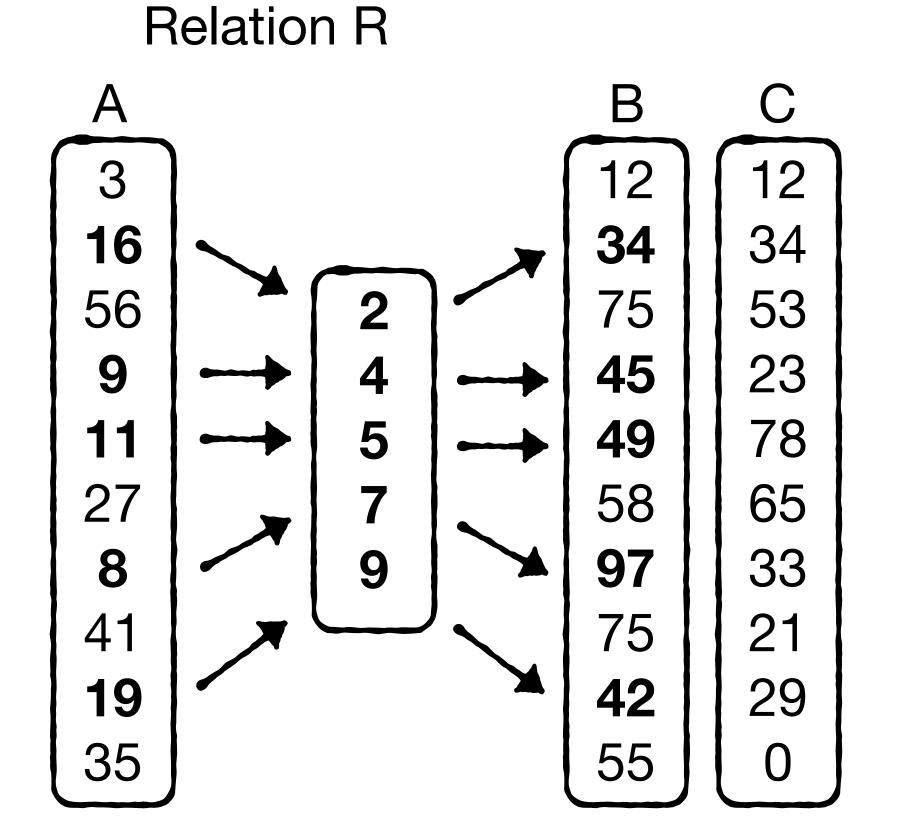
So in this regime, we may get a compression rate smaller than 2

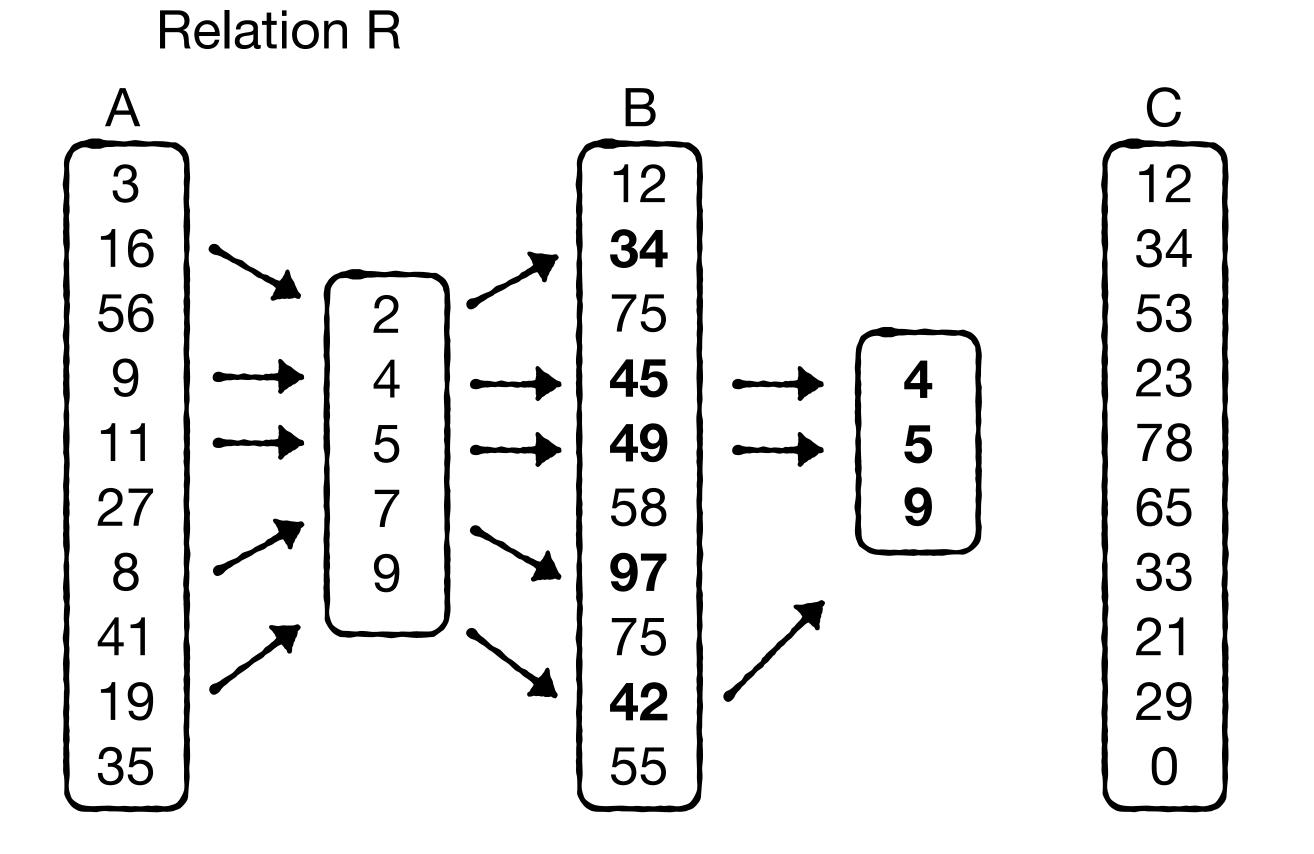
At the same time, the dictionary becomes large and may exceed the size of the CPU caches, making decompression more expensive

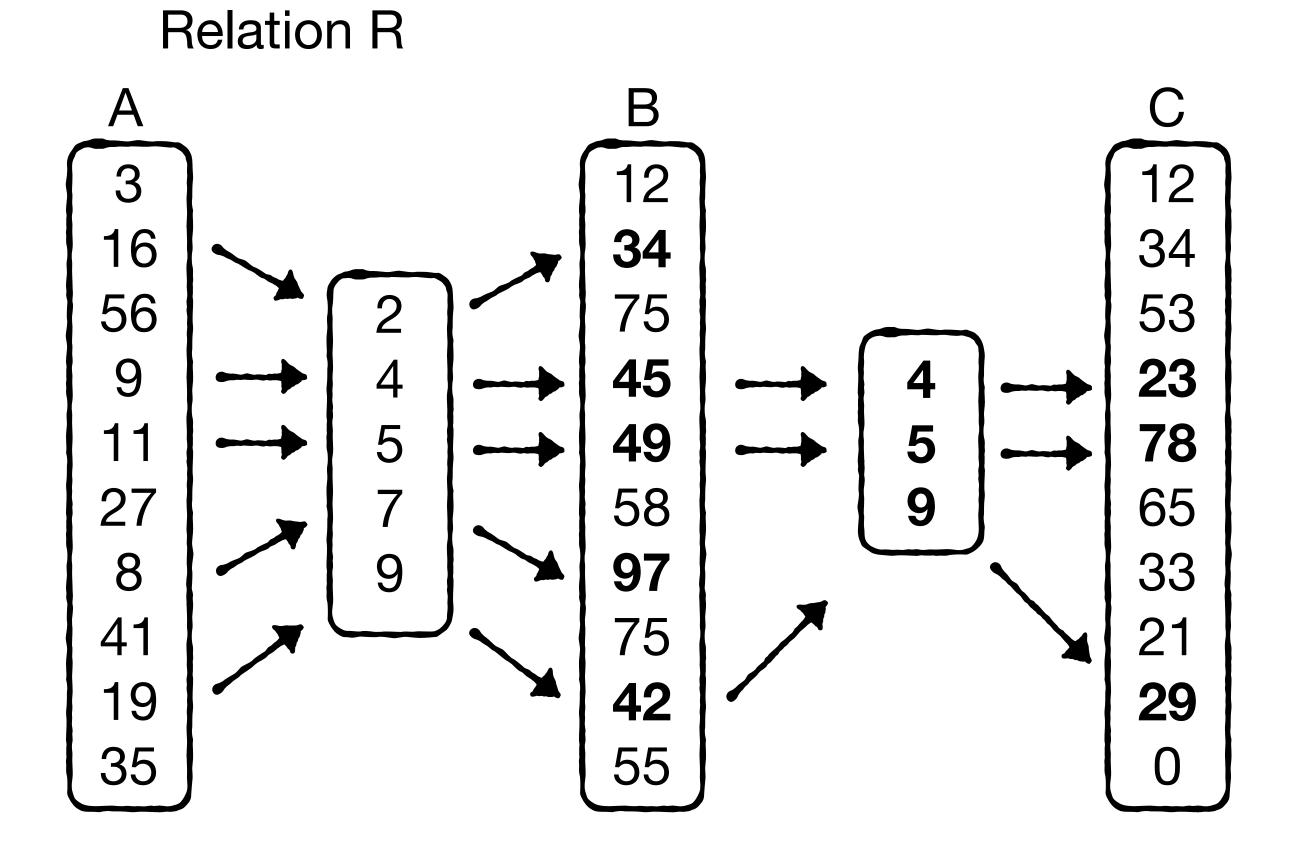
Describe how to physically process the following query using late materialization. No information on cardinality is provided.

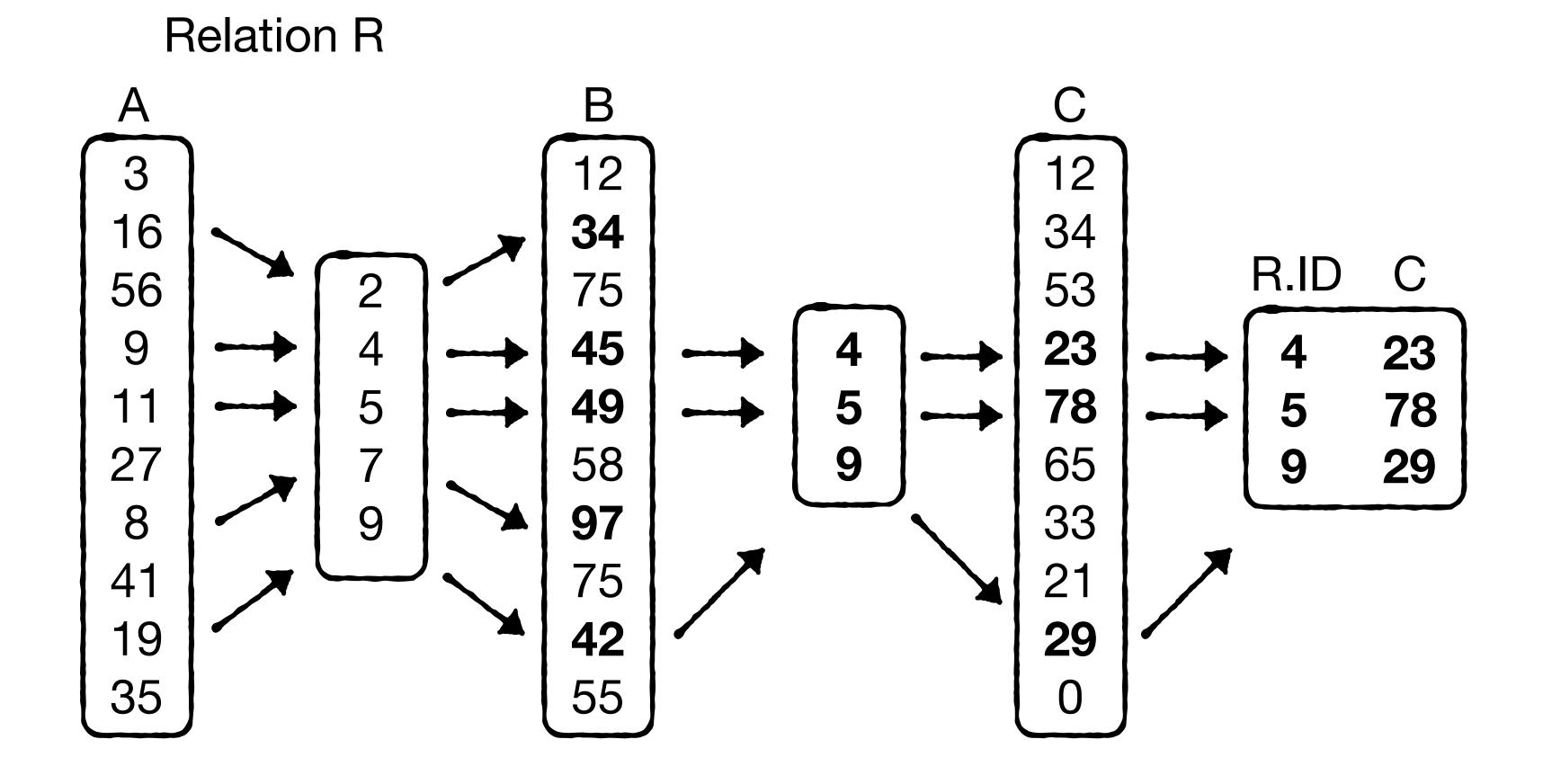




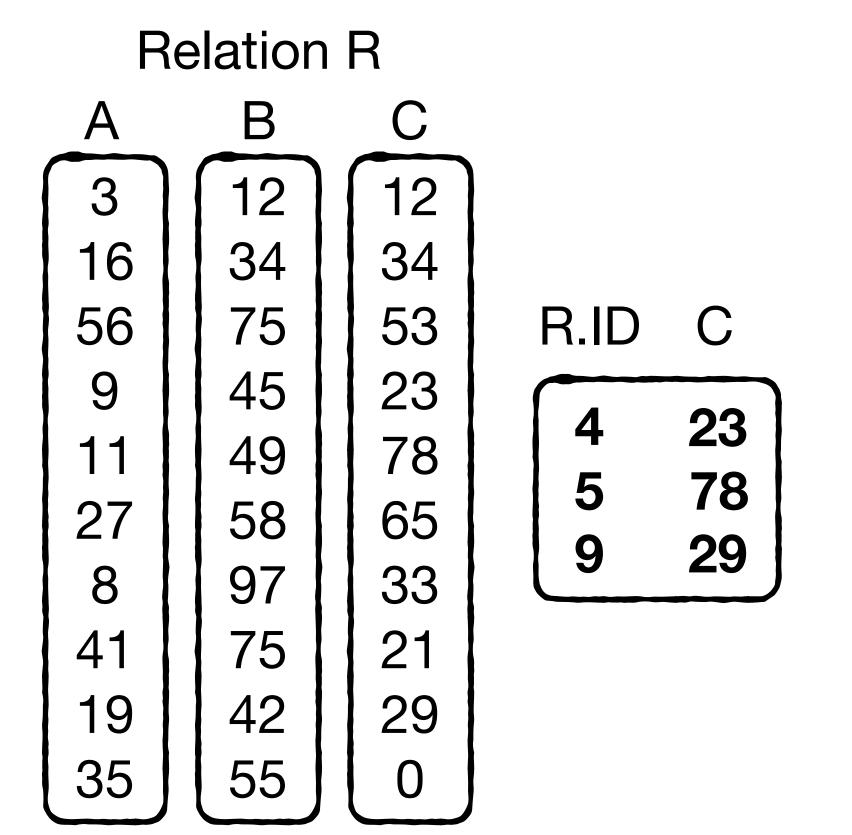






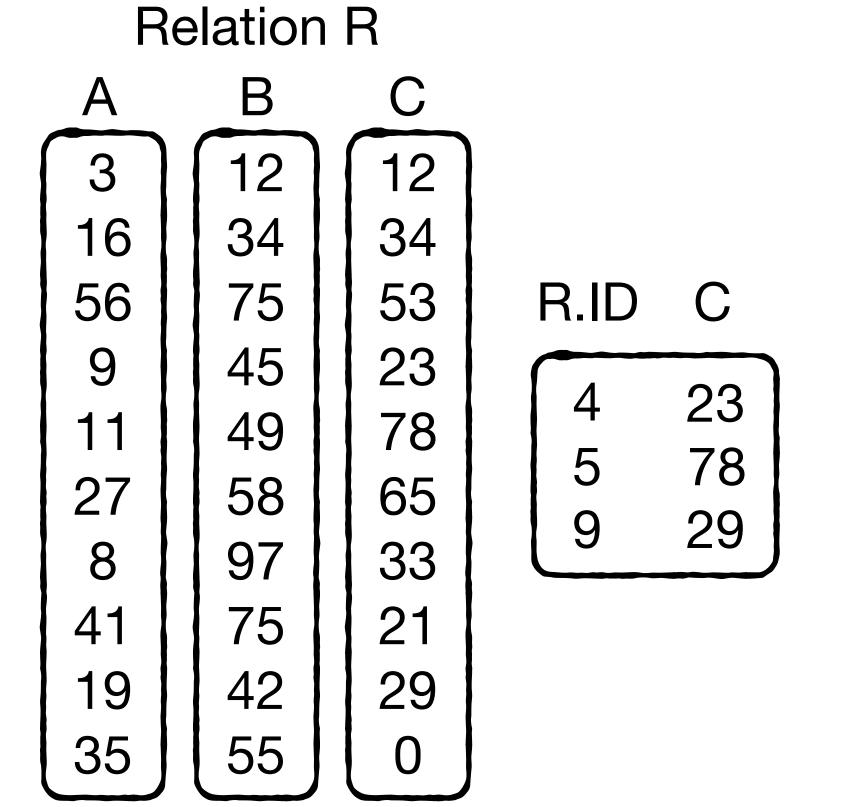


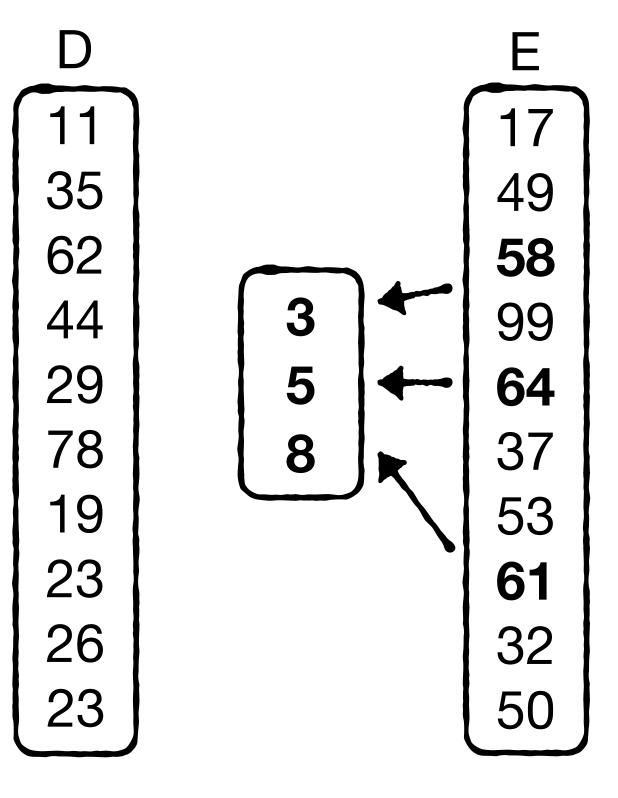
Select sum(A) from R, S where C=D and 5<A<20 and 40<B<50 and 55<E<65

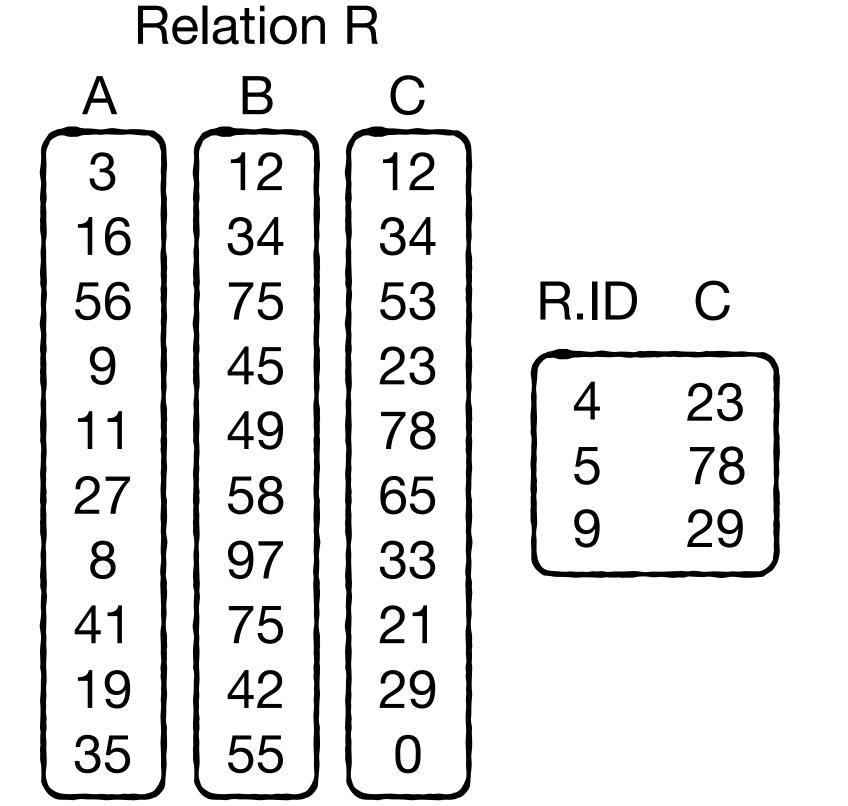


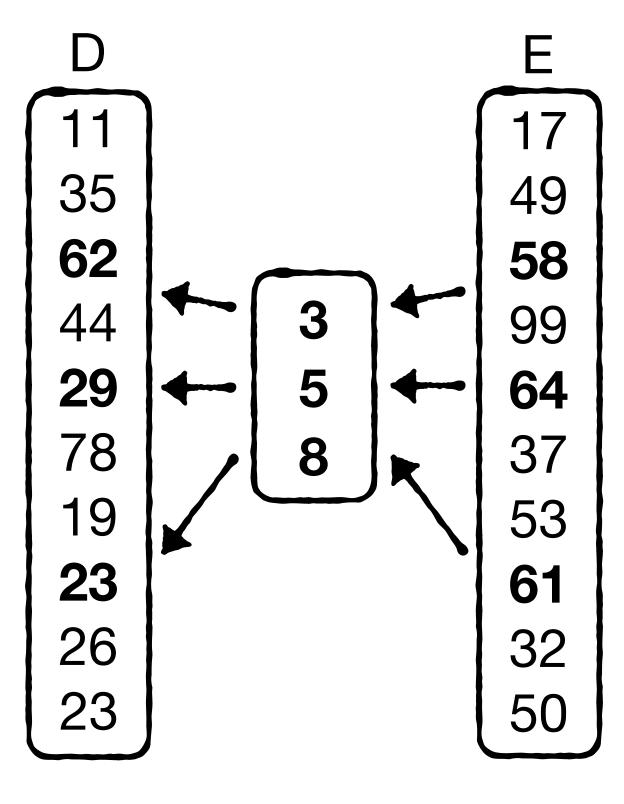
#### Relation S

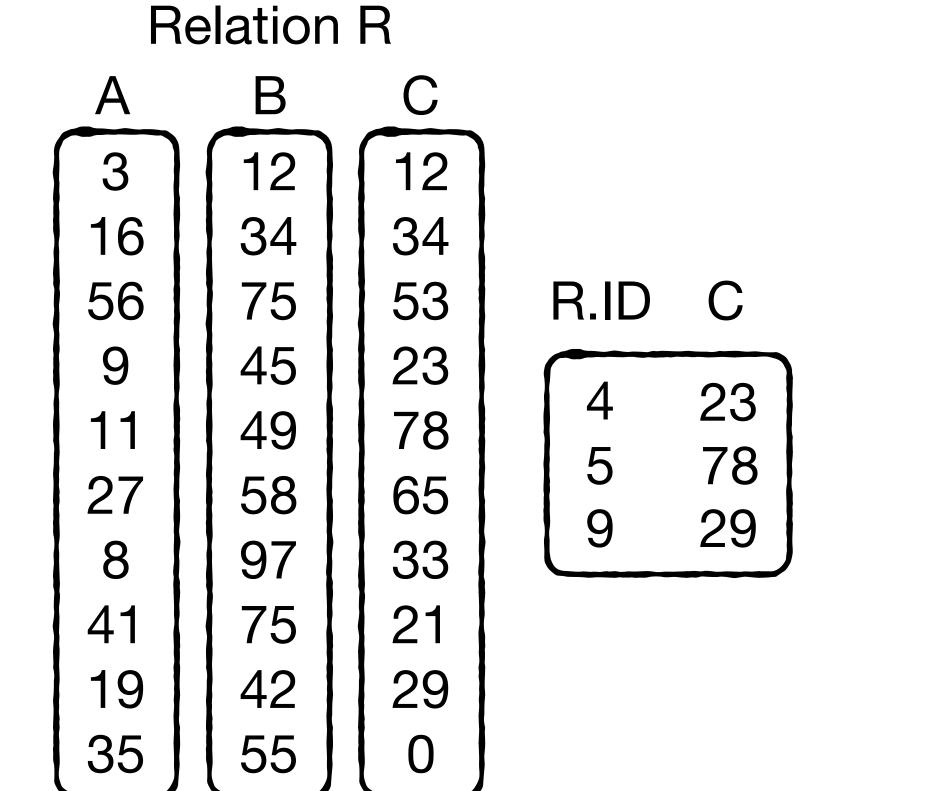
$\overline{D}$	E
11	17
35	49
62	58
44	99
29	64
78	37
19	53
23	61
26	32
23	50

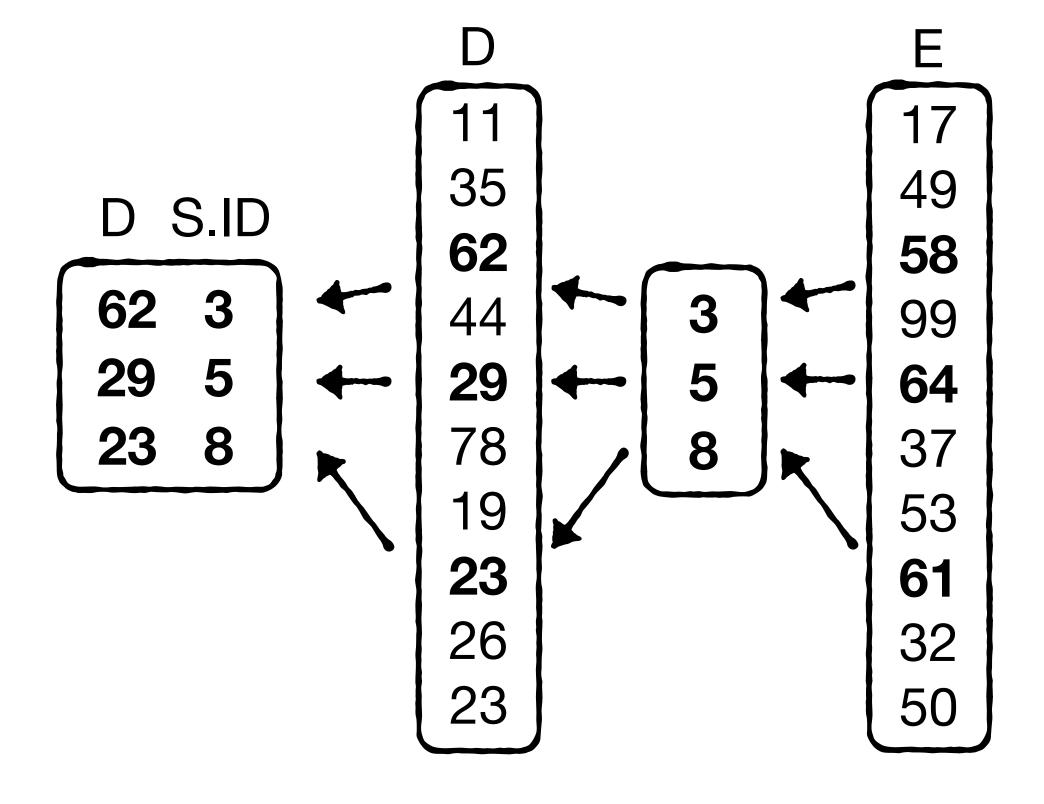






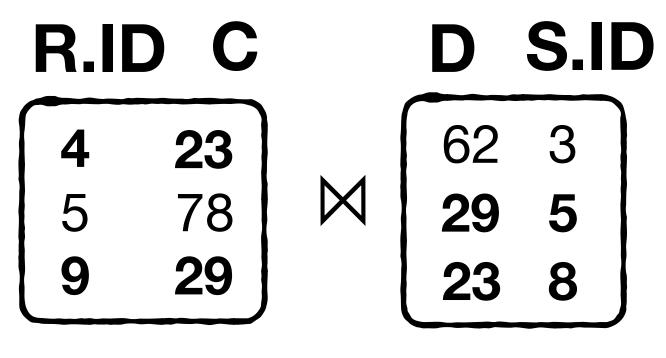


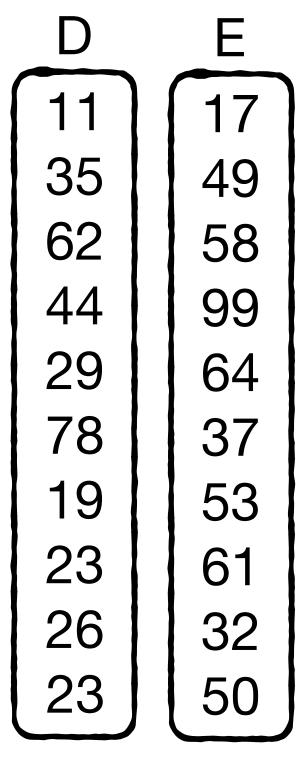


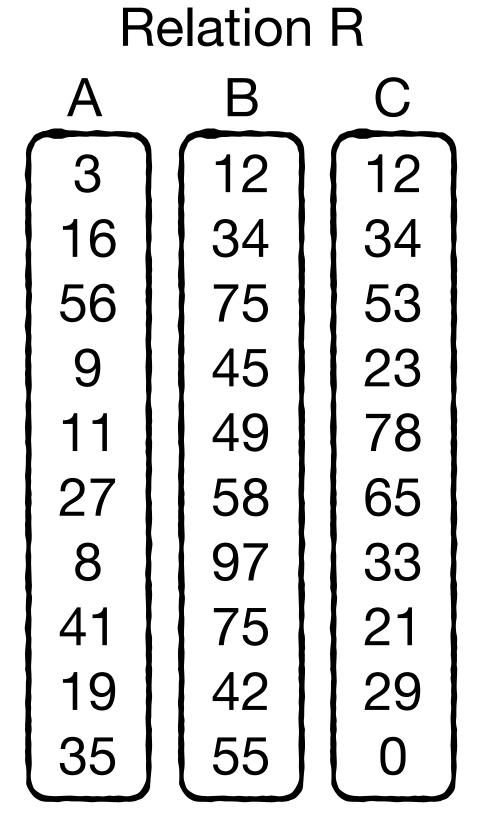


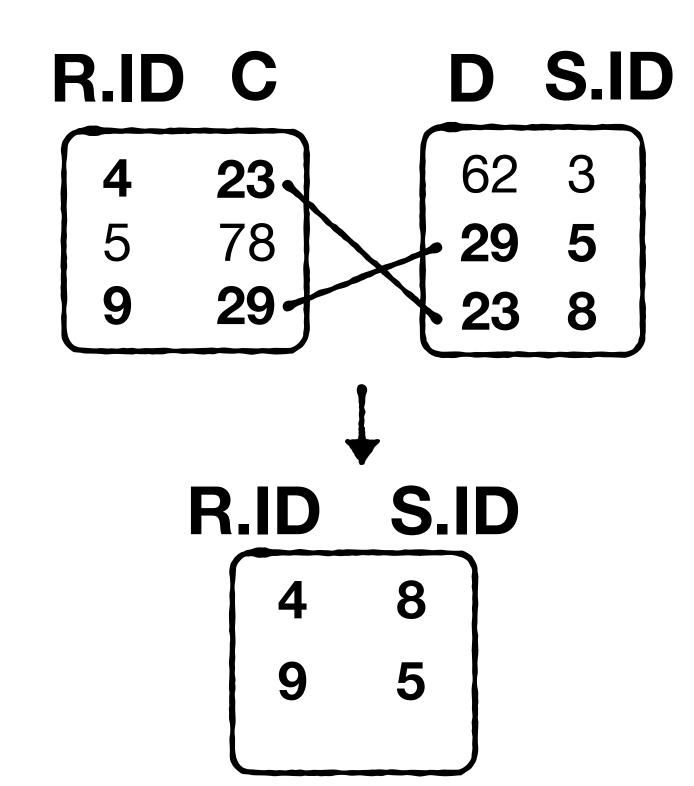
Select sum(A) from R, S where **C=D** and 5<A<20 and 40<B<50 and 55<E<65

# Relation R B









D	E
11	17
35	49
62	58
44	99
29	64
78	37
19	53
23	61
26	32
23	50

Select **sum(A)** from R, S where C=D and 5<A<20 and 40<B<50 and 55<E<65

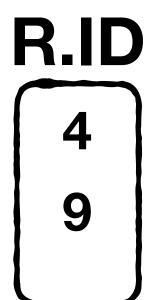
1935

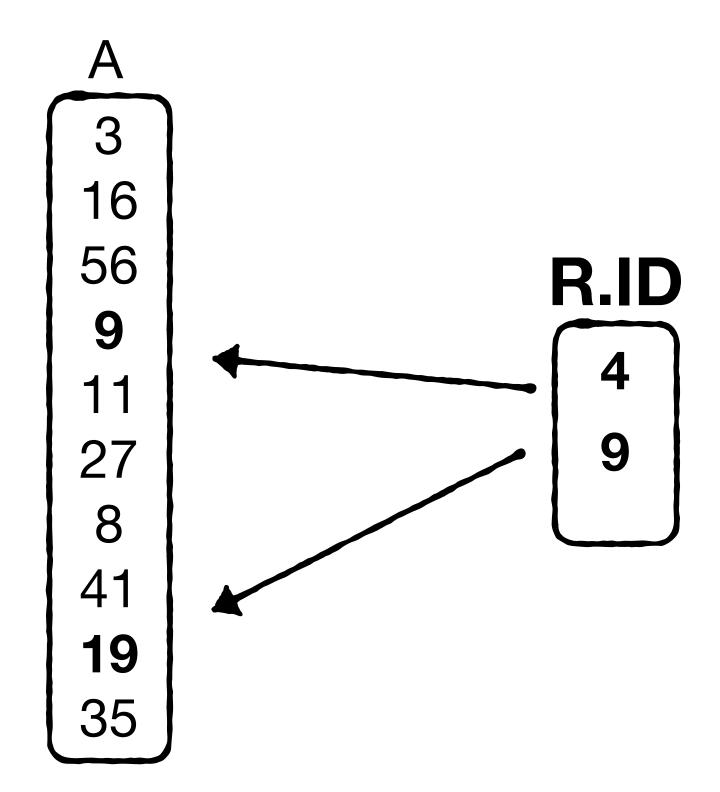
R.ID S.ID

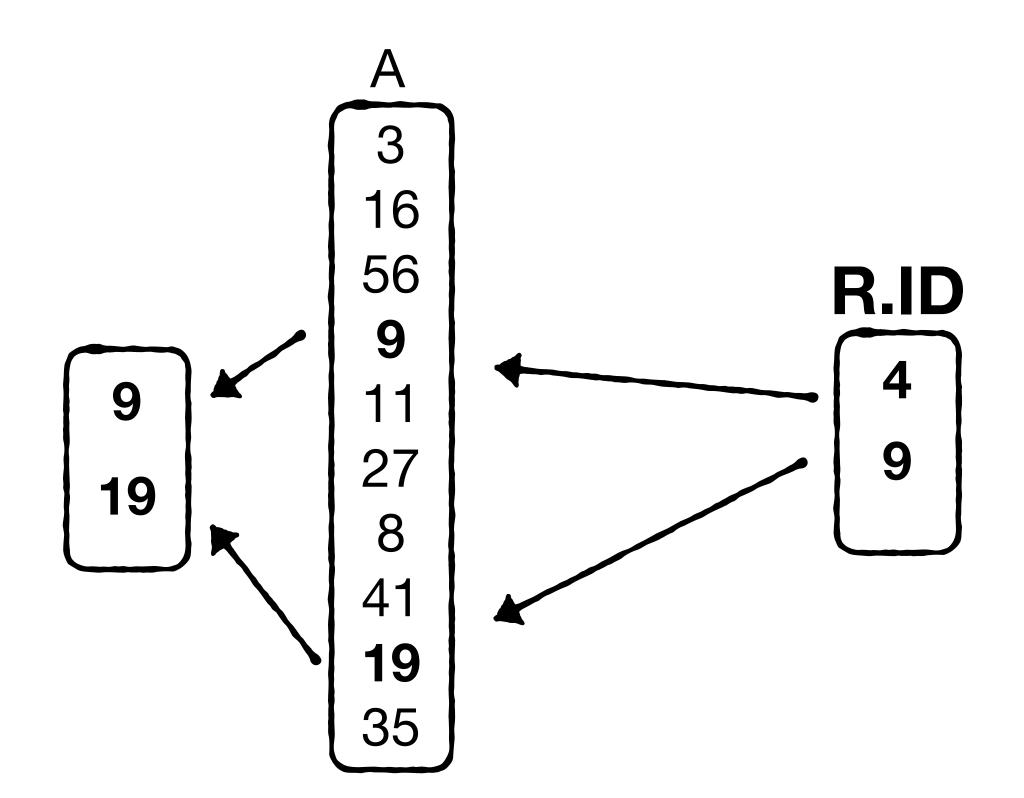
4 8
9 5

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# Relation R







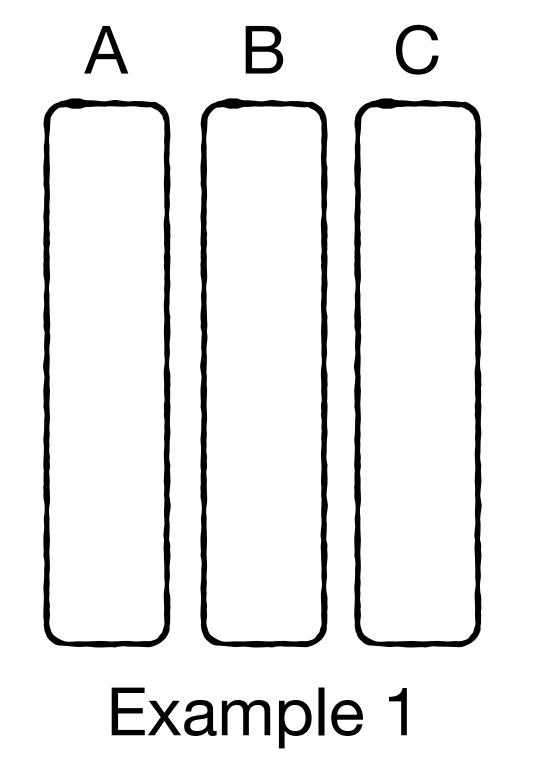
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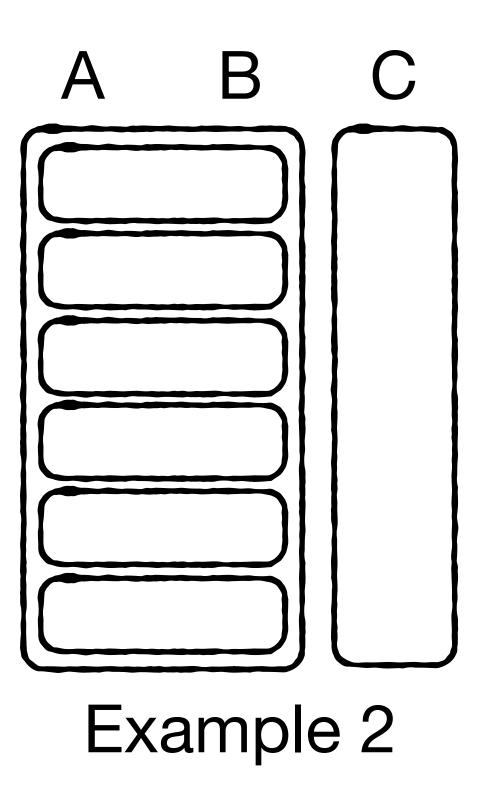
9 +

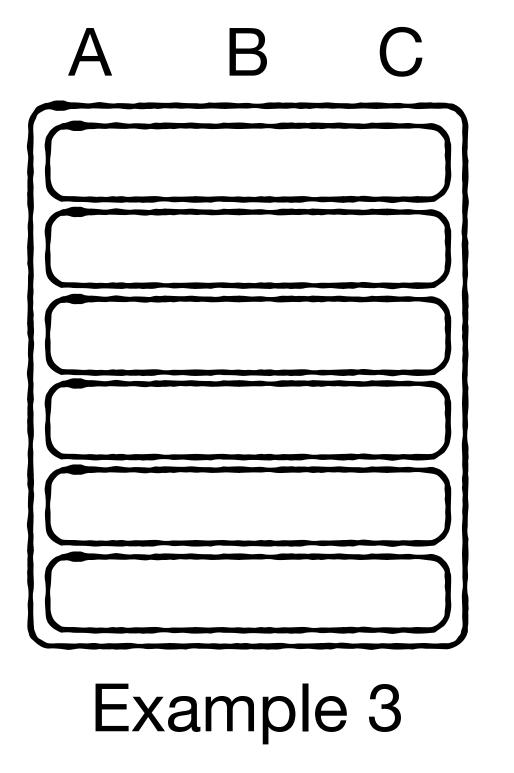
19

= 28

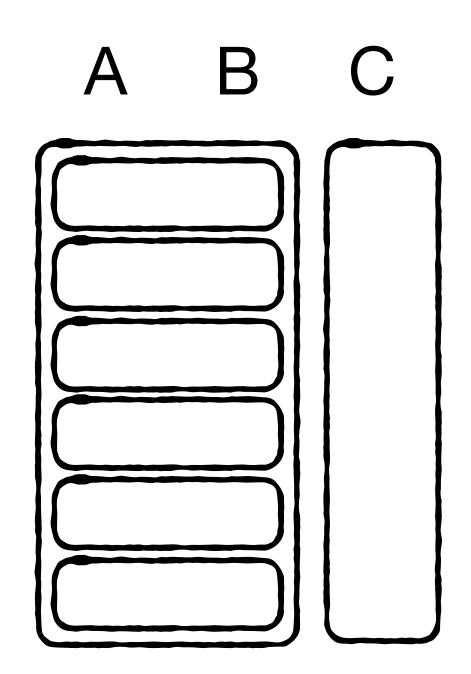
"Column groupings" are a way to store values from across several columns in a table in a row-store format. In which cases is it beneficial to group columns, and in which cases is it harmful?







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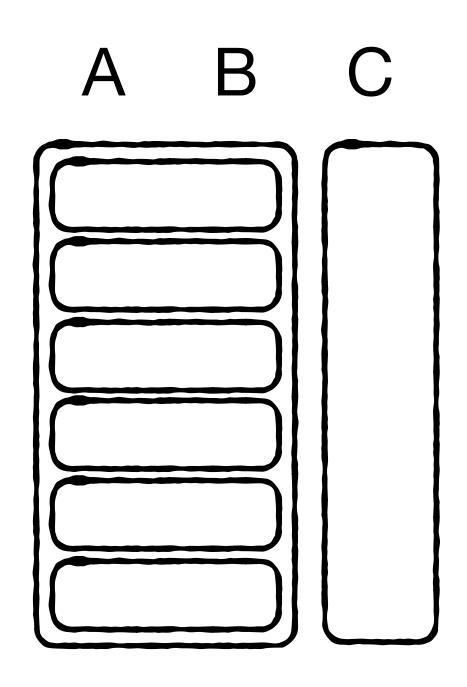


Good when values of corresponding rows along grouped columns are always accessed at the same time

#### **Example:**

Select A, B, C where C=""

"Column groupings" are a way to store values from across several columns in a table in a row-store format. In which cases is it beneficial to group columns, and in which cases is it harmful?



harmful for queries that access or filter based on only a subset of columns in a group

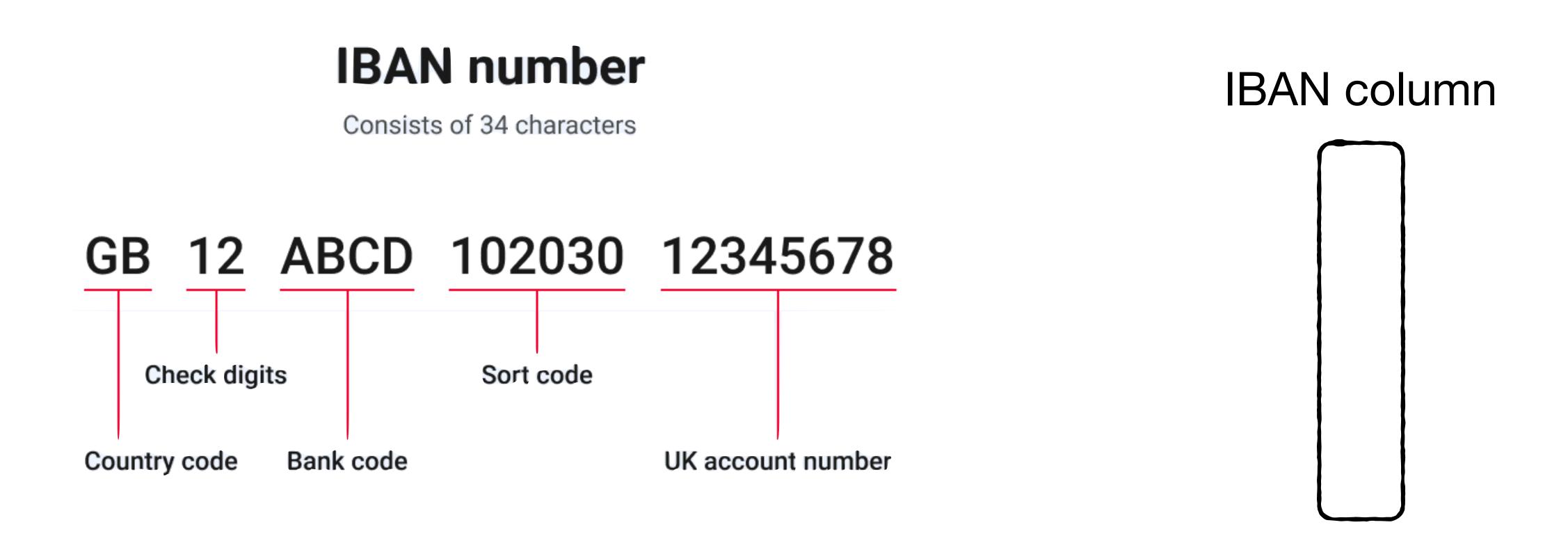
#### **Example:**

Select A, C

Select B, C where B=""

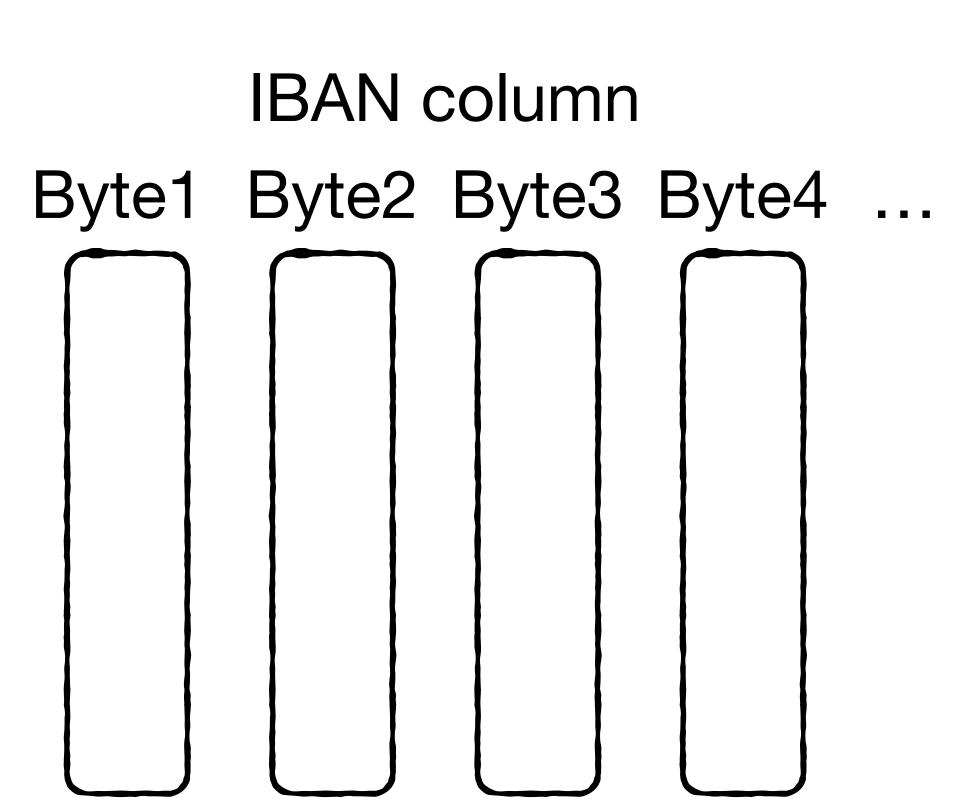
Select A, B where A = "..."

Consider a column of International Bank Account Numbers (IBANs). We have queries of the form "select \* where IBAN = x". How can we optimize for such queries without sorting the column or using an index? Propose a generic solution that's suitable beyond just IBANs.



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Traverse columns from most selective byte first and employ late materialization.

