TERTIARY STORAGE ORGANIZATION FOR LARGE MULTIDIMENSIONAL DATASETS

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Introduction

- Large multidimensional datasets are found in data warehousing, satellite data processing, high energy physics
 ...
- ullet Large amounts of data (> 1 terabyte), stored in tertiary storage
- Data permutation problem
 - Data exists/generated in native order
 - To be converted into storage order
 - Limited permutation space

Data permutation by multiple passes over input data



A two dimensional 16 element dataset

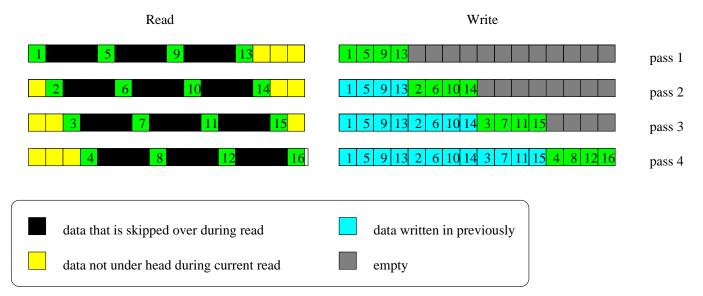
generated in native order

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16

can be transformed into a storage order

1 5 9 13 2 6 10 14 3 7 11 15 4 8 12 16

by making 4 passes over input dataset





1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16

A two dimensional 16 element dataset

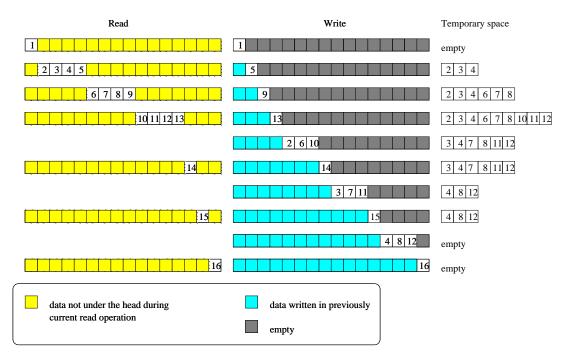
generated in native order

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16

can be transformed into a storage order

1 5 9 13 2 6 10 14 3 7 11 15 4 8 12 16

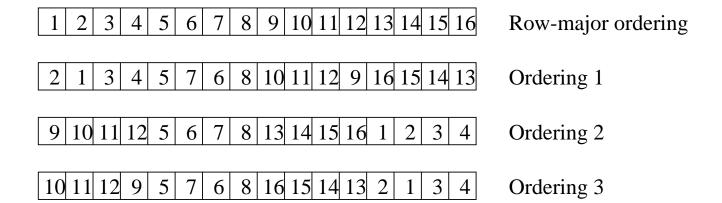
by using temporary space holding not more than 9 data items

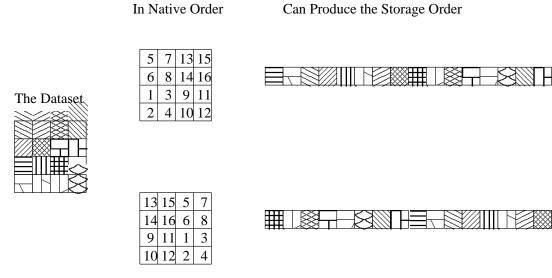


Same access pattern, more than one suitable storage pattern

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16

A two dimensional 16 element dataset





Different native orders for the same access pattern can mean different storage orders under constrained permutation space

Computing Affinities

Given two data items in the dataset, I_i and I_j , and a set of expected query types and their execution probabilities, we define $p(I_i, I_j)$ as the *affinity* between I_i and I_j . Affinity between two data items is the probability that if I_i is accessed by a query then I_j is also accessed and can be compute as:

```
\sum_{Q_i=Q_1}^{Q_k}(p_i\times\prod_{D_j=D_1}^{D_n}\lambda_i^j) where, Q_i\quad i^{th} \text{ query type} p_i\quad \text{execution probability of }Q_i D_j\quad j^{th} \text{ dimension} \lambda_i^j\quad =\quad 1\quad \text{when }Q_i(j)=ALL \quad =\quad 1\quad \text{when }Q_i(j)=ANY \text{ and }I_1(j)=I_2(j) \quad =\quad 1\quad \text{when }Q_i(j)=VALUE \text{ and }I_1(j)=I_2(j)=v_i^j \quad =\quad 1\quad \text{when }Q_i(j)=RANGE \text{ and }|I_1(j)-I_2(j)|\leq r_i^j \quad =\quad 0\quad \text{otherwise} Q_i(j)\quad \text{selection criteria of }Q_i \text{ in }D_j I_1(j)\quad \text{coordinate of }I_1 \text{ in }D_j I_2(j)\quad \text{coordinate of }I_2 \text{ in }D_j v_i^j\quad \text{value parameter for }Q_i \text{ from the domain of }D_j r_i^j\quad \text{range parameter for }Q_i \text{ from the domain of }D_j
```

Greedy Heuristic

```
while (there are more data items)
  if (there is space left in the temporary storage)
    input "some" data items
  endif
  if (there are data items in the temporary storage)
    output "some" data items
  endif
endwhile
```

- ullet If there are $d \leq D$ pages free in the temporary storage, how many data items of size S pages to input in each iteration?
- If there are k data items $(k \times S \leq D)$ in the temporary storage, how many and which of the k data items to output in each iteration?

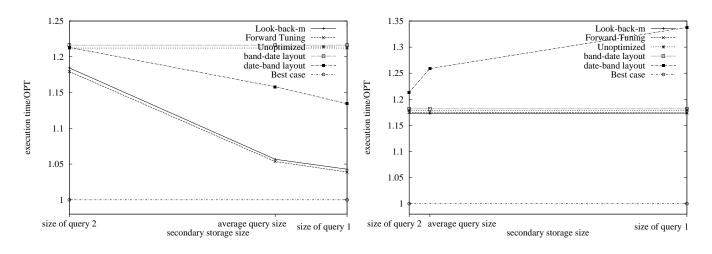
Algorithms

- ullet Look-back m : remember last m data items
- ullet Forward tuning : Look-back m with ties resolved
- Date-Band, Band-Date : fixed order strategies

Workloads

- Workload 1: native order differs substantially from storage order
- Workload 2 : native order differs very little from storage order

Performance Results



Workload 1

Workload 2