

The index merge scheduling problem

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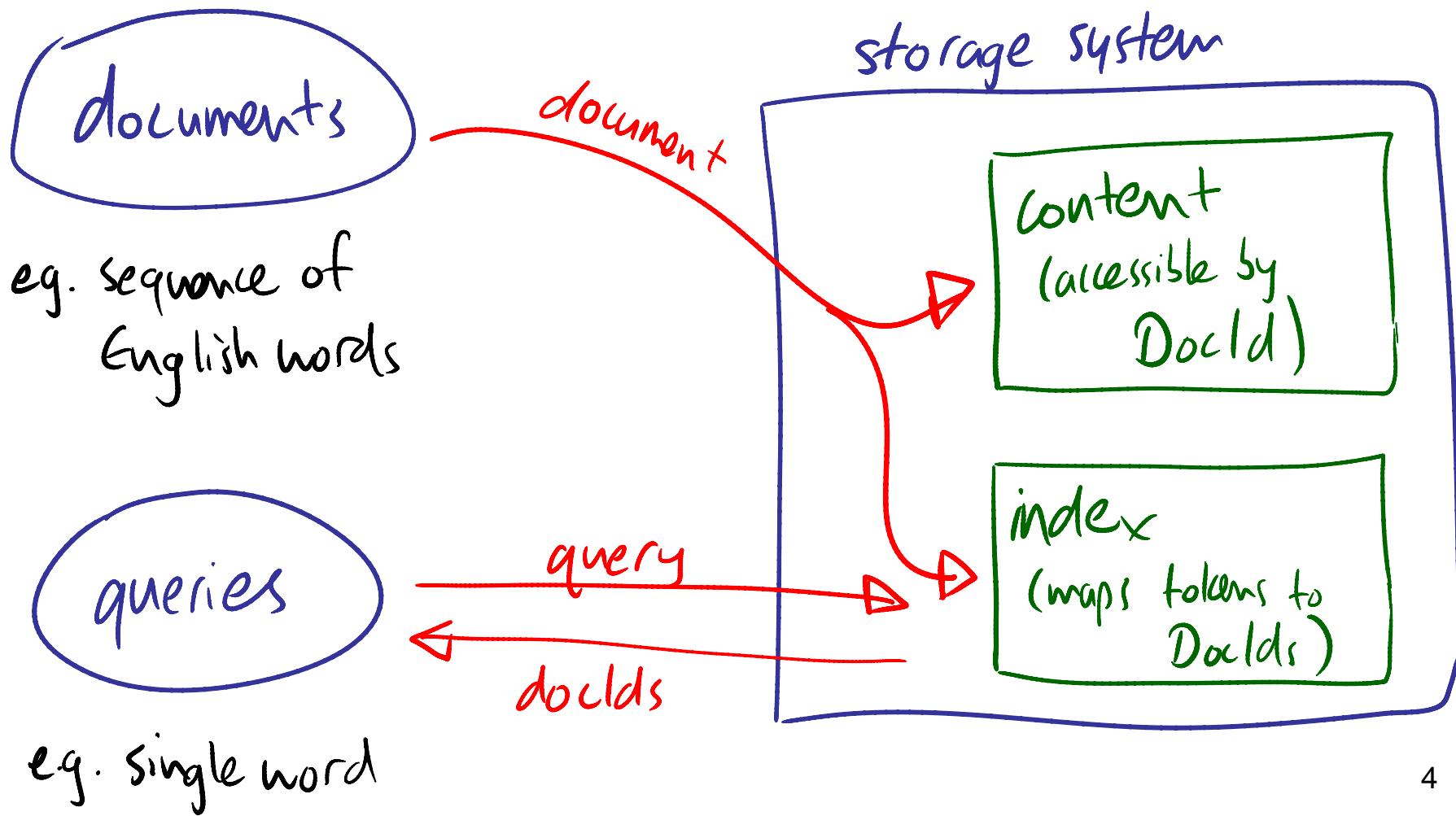
Motivation: full text indexing of dynamic content

- data arrives continuously
- queries must reflect latest arrivals
- examples:
 - webmail (Yahoo Mail, Gmail, Hotmail,...)
 - blogs (MSN spaces)
 - news stories (Google News)
 - desktop search
 - Web search

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- fous on this
as main
example*

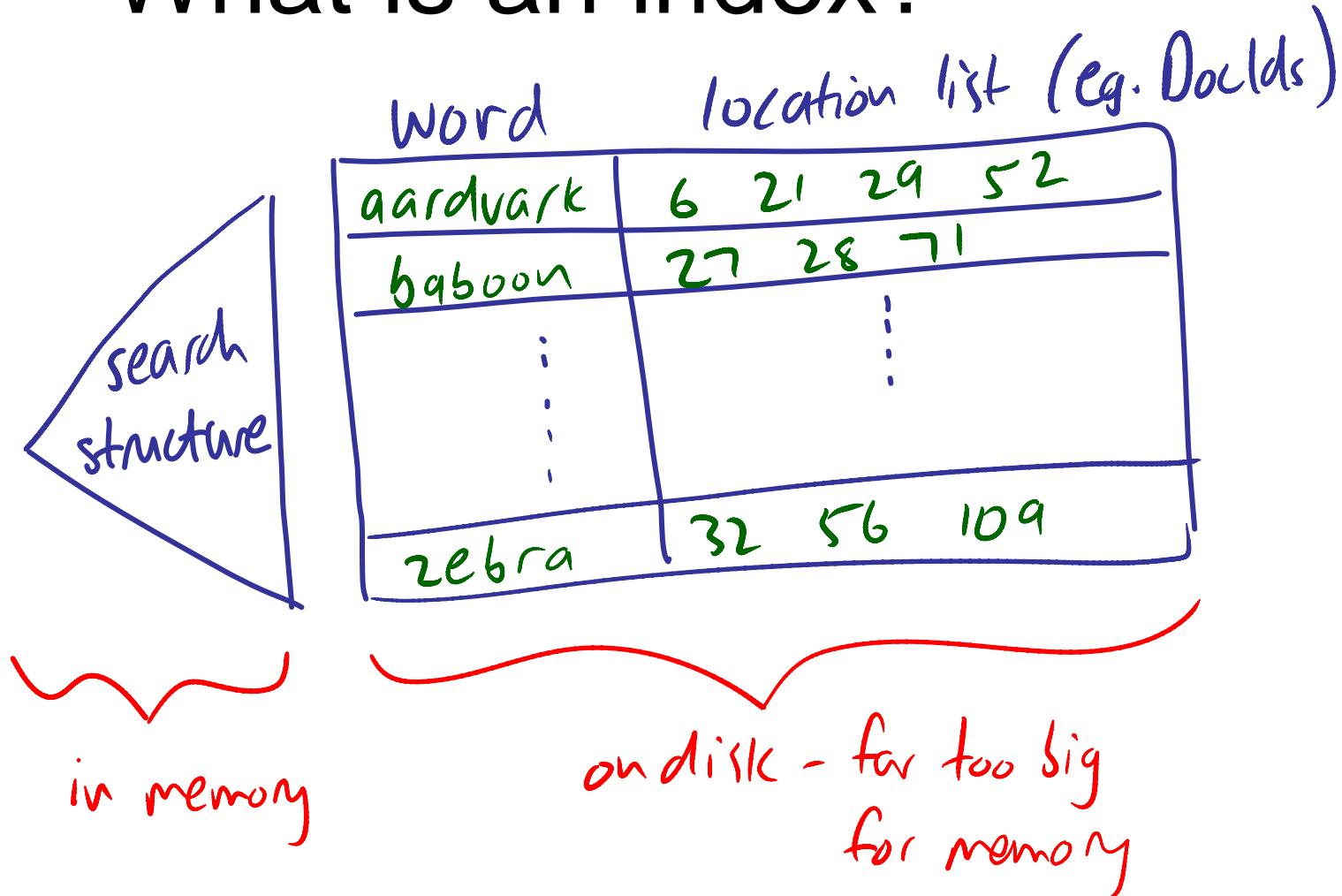
High -level overview of dynamic content indexing



the index merge scheduling problem: roadmap

1. single index is insufficient for dynamic content – need multiple indexes, and therefore need occasional index merges

What is an index?



What is an index?

Logically :

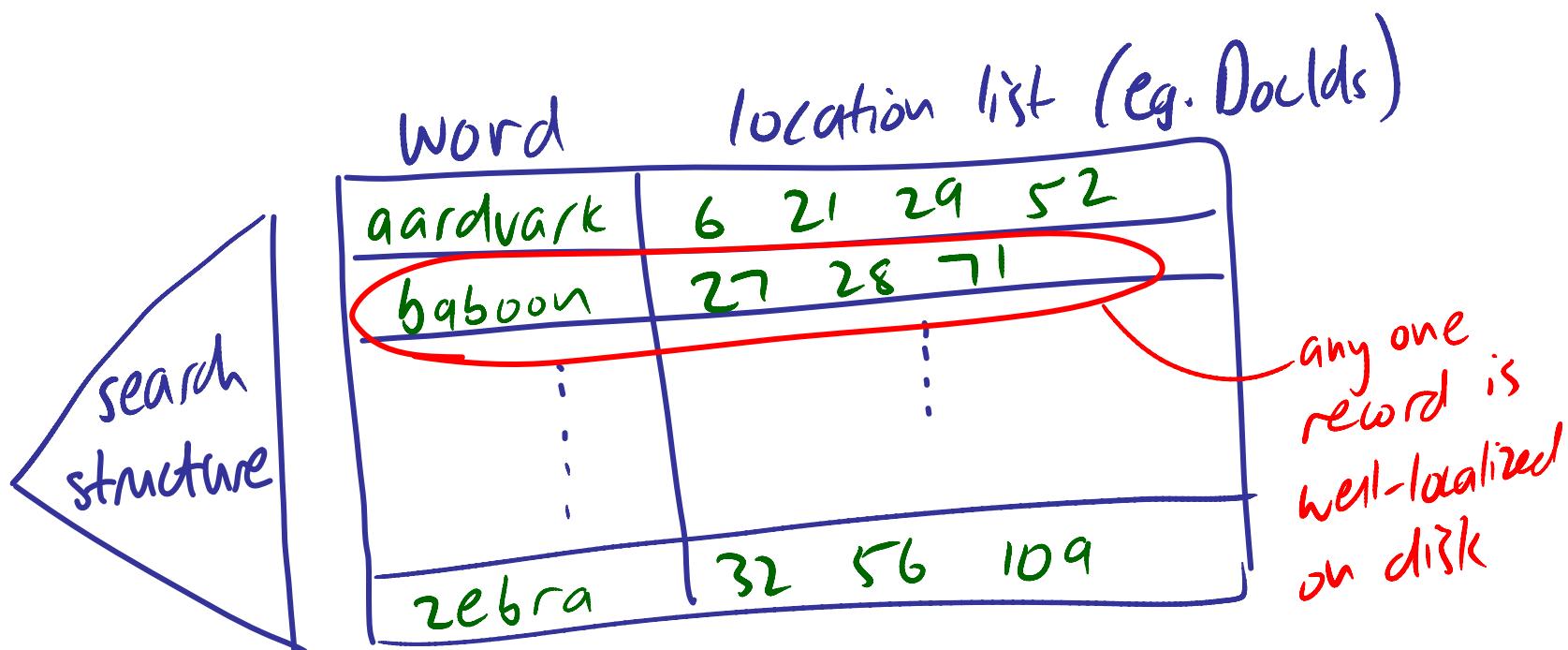


Word	location list (eg. DocIds)
aardvark	6 21 29 52
baboon	27 28 71
:	:
zebra	32 56 109

Physically : B-tree, vanilla inverted file,
fancy inverted file

(Zobel et al 93, Lomax 88, Carey et al 89)

Definition: a *single index* can retrieve the entire location list for a word in a single I/O

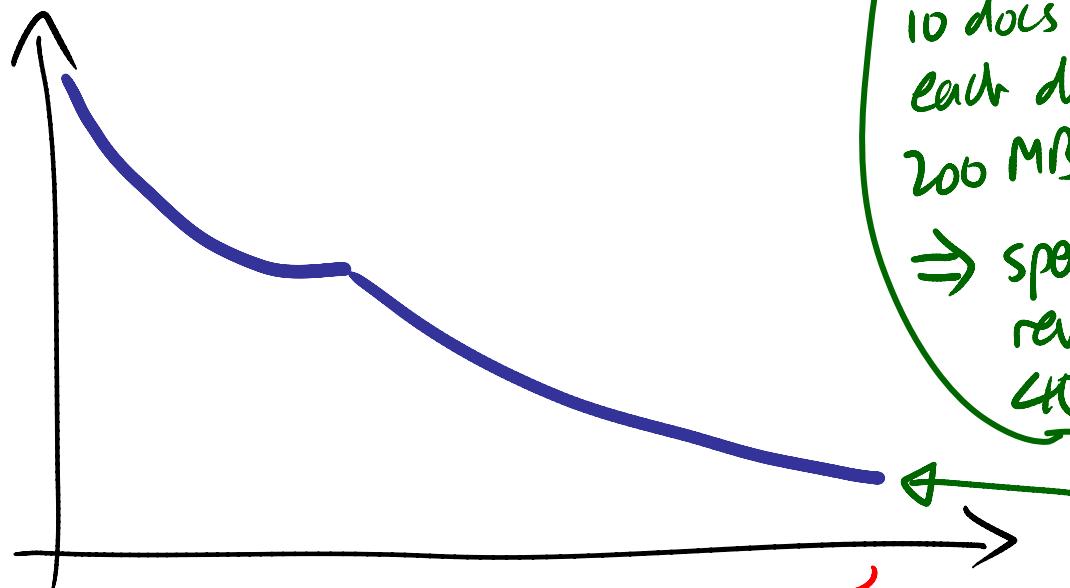


Naïvely indexing dynamic content is far too slow

- Recall:
 - random I/O is slow (10 ms seek time per I/O)
 - sequential I/O is fast (1 seek + 100 MB/s)
- Example of naïve indexing: scientific paper
 - 100 kB as text file
 - contains 2500 unique indexable words
 - naïvely updating each index entry would take 2500 times longer than writing the file sequentially!

Claim: a single index is too slow

amortized
I/O time for
indexing 1
document



400 GB index
10 docs/sec
each doc 10 kB
200 MB doc cache
⇒ spend 20 mins
rewriting, every
40 mins

memory size
for document cache

update individual
location lists via
random I/O

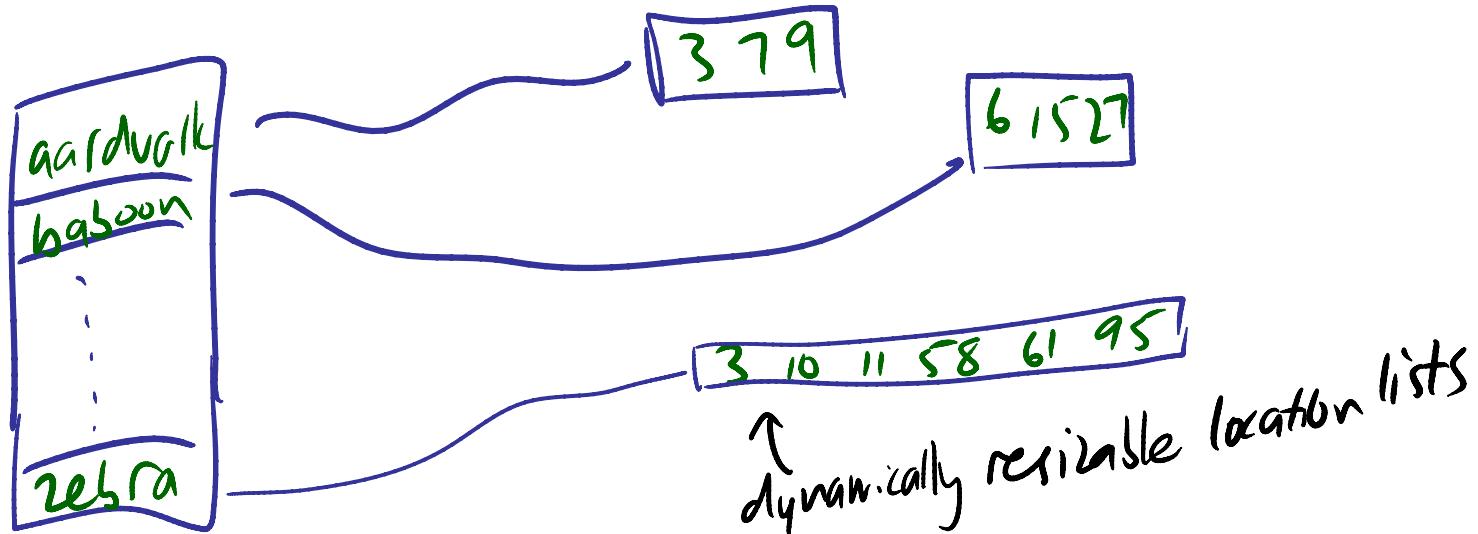
rewrite whole index
sequentially

Therefore, we need multiple indexes

- 2 basic types: in-memory and on-disk

In-memory index

can easily add new docs



on-disk index file

can't add new docs



Note Easy to write out in-mem index to an index file, sequentially¹¹

Index files can be merged using only sequential I/O and little memory

index file 1



index file 2



merge of 1 & 2



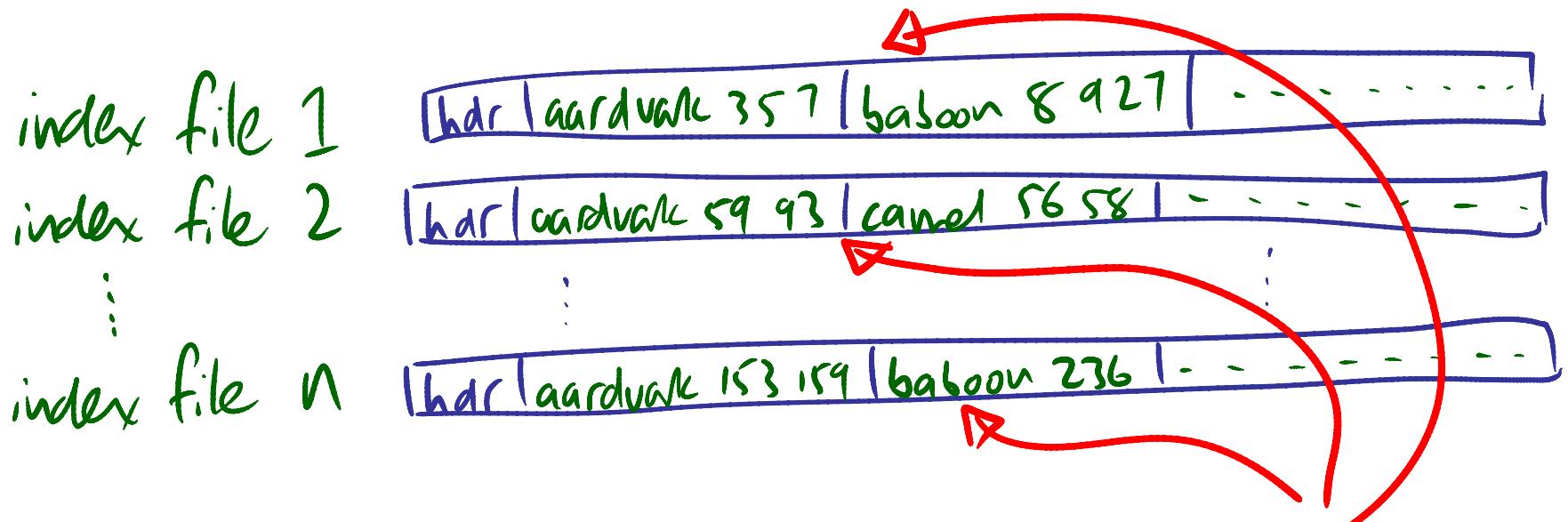
I/O cost of merge = cost of sequential read of 1+2
+ sequential write of 1+2

- can merge many indexes into 1 simultaneously
- in-memory indexes can be merged with on-disk index files
- writing out in-memory index to disk by itself can be regarded as a trivial merge

Basic strategy for dynamic indexing: accumulate and merge

- Repeat:
 - accumulate as many documents as possible in an in-memory index
 - merge in-memory index with zero or more index files
- Optionally, in parallel, repeat:
 - merge some index files

Why merge? Because query cost is proportional to number of indexes



Exception: if index files are sorted by relevance,
some queries require less I/O

the index merge scheduling problem: roadmap

1. single index is insufficient for dynamic content – need multiple indexes, and therefore need occasional index merges
2. scheduling merges is related to the cost-distance problem in network construction

Formal definition of index merge scheduling problem

given finite sequence of events $(e_1, e_2, \dots, e_T) \in \{D, Q\}^T$

\nearrow \nearrow
data arrival query arrival

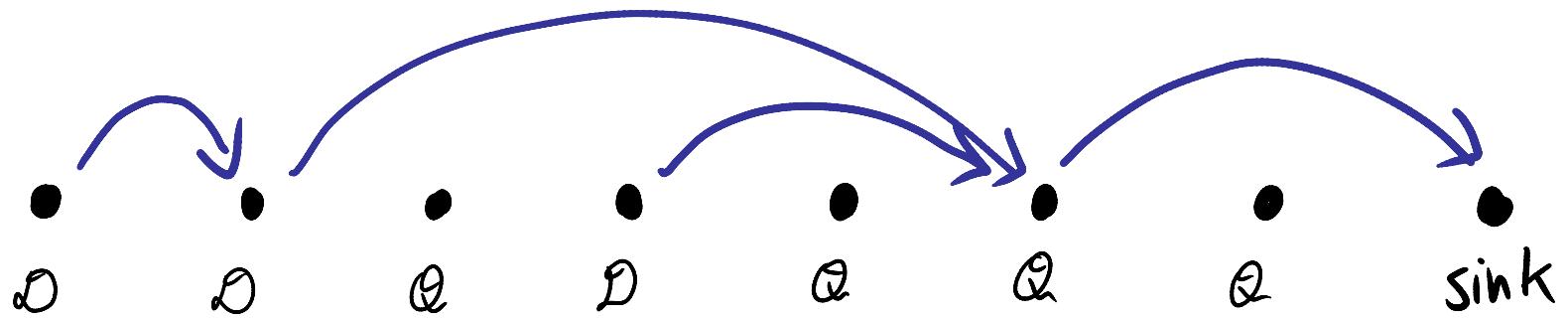
construct merge schedule

$m : \{1, 2, \dots, T\} \rightarrow$ set of indexes
to be merged at
the t

to minimize total cost

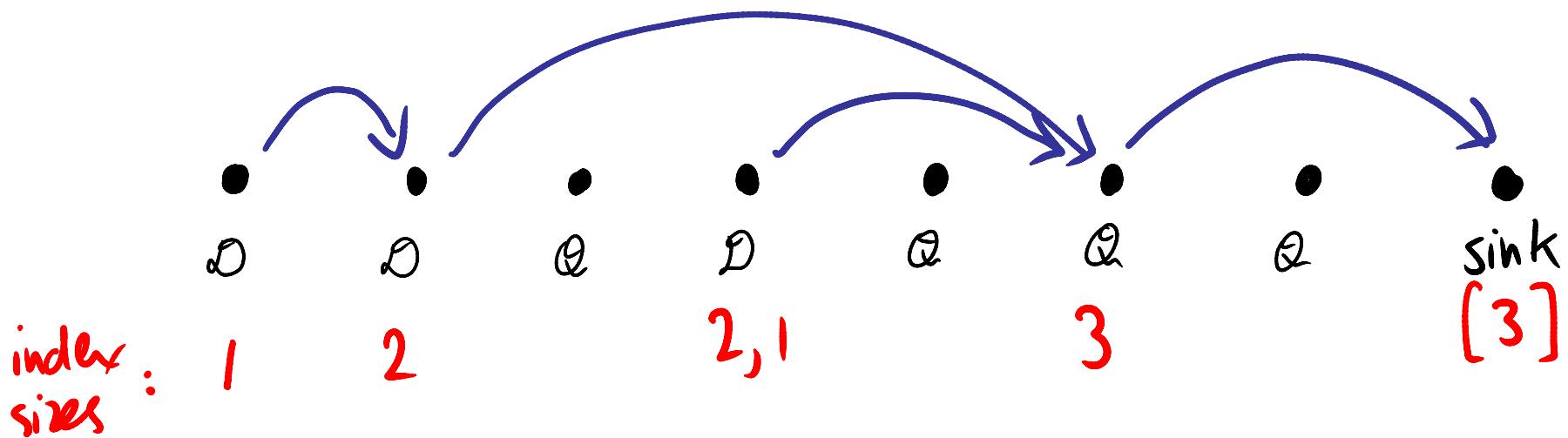
$$C = C_{\text{merge}}(T) + C_{\text{query}}(T)$$

merge scheduling can be regarded as constructing minimum-cost network on a certain graph



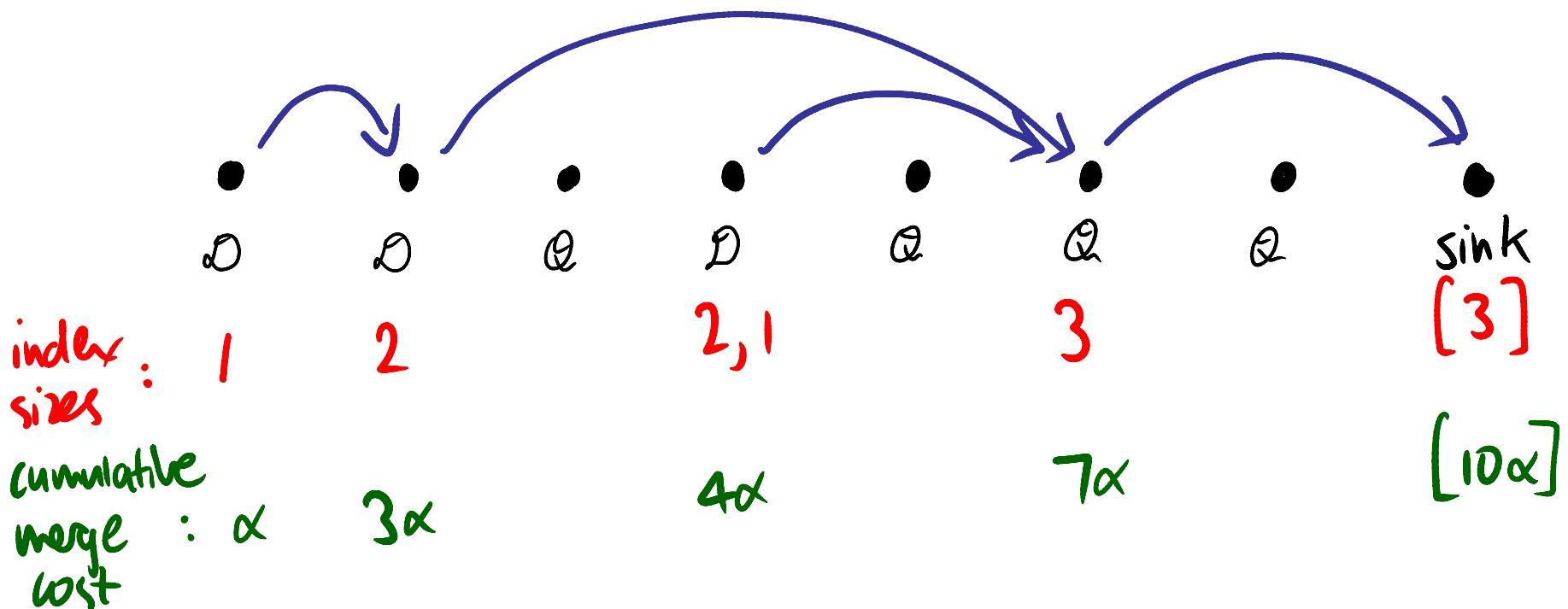
edge from A to B means “take all data written immediately after event A and merge it immediately after event B”

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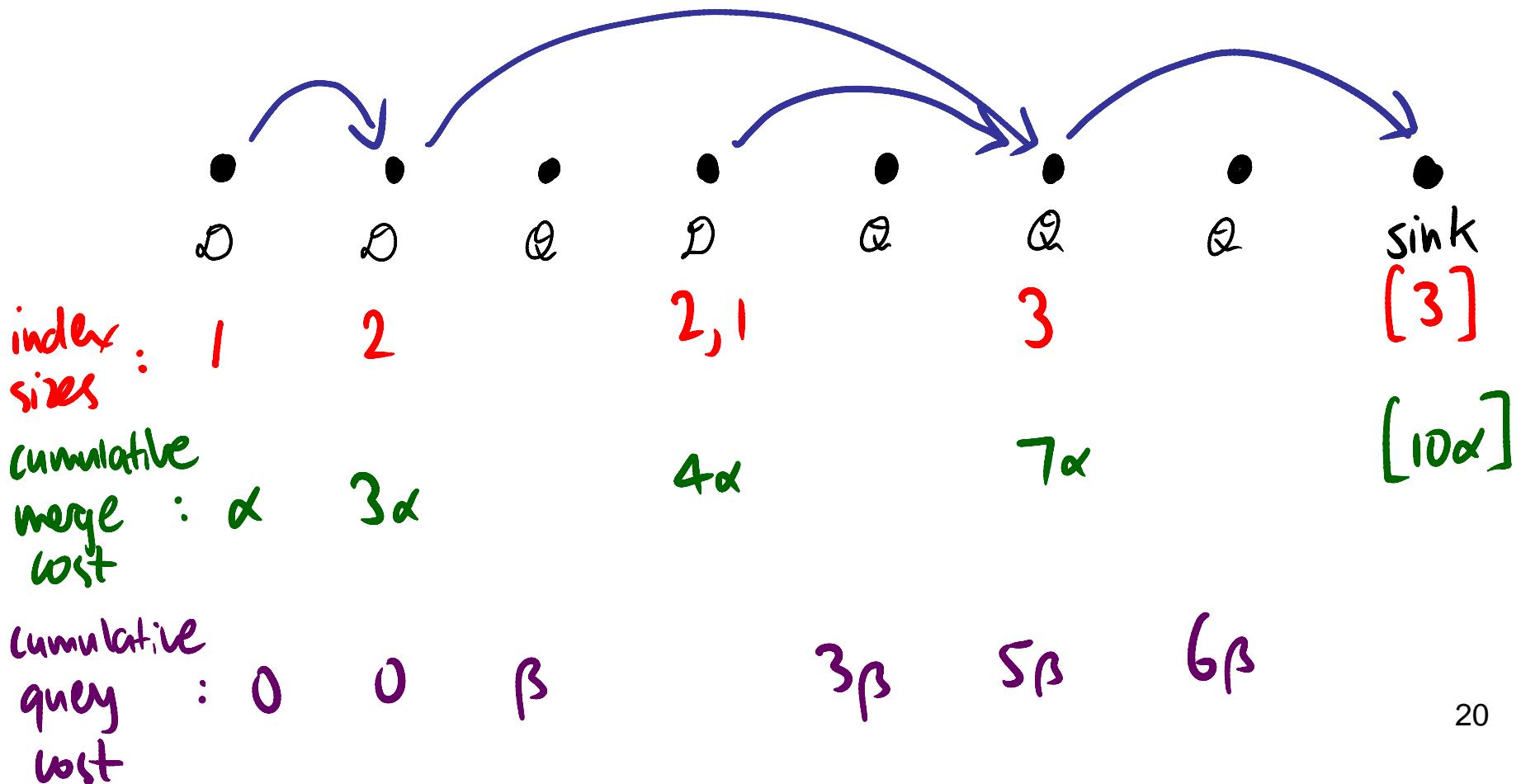


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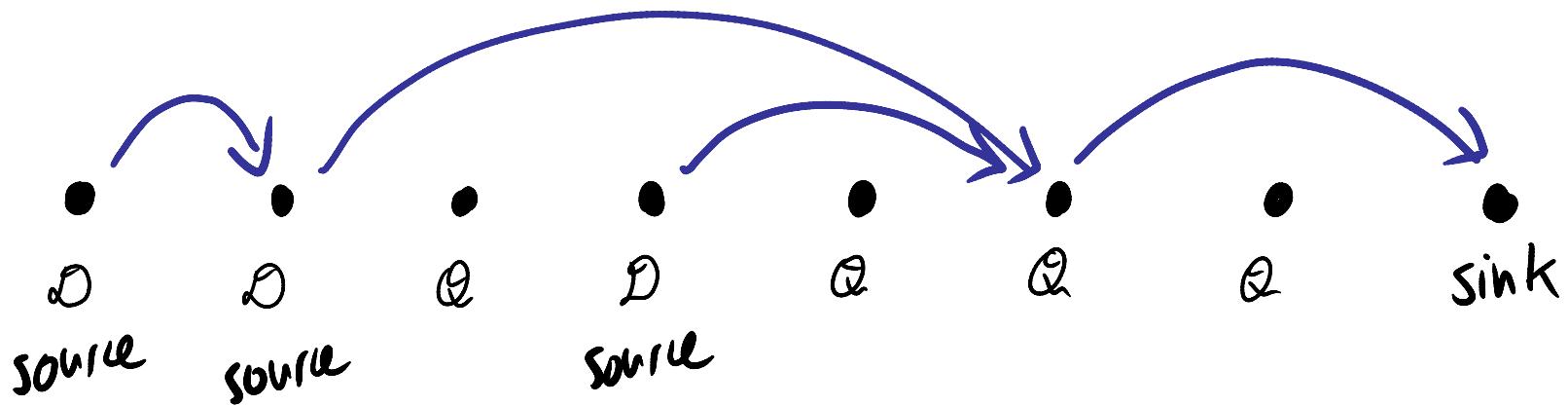
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merge scheduling can be regarded as constructing minimum-cost network on a certain graph



merge cost is sum of path lengths
from sources to sink

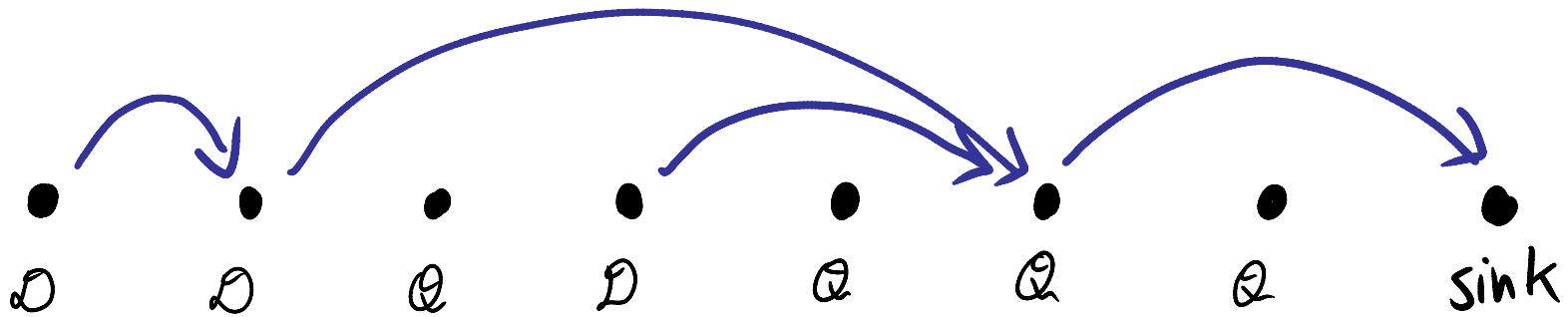


$$C_{\text{merge}} = \alpha \sum_{\text{sources } s} \text{len}(s, \text{sink})$$

merge price

path length
i.e. number of hops

query cost is sum of edge costs

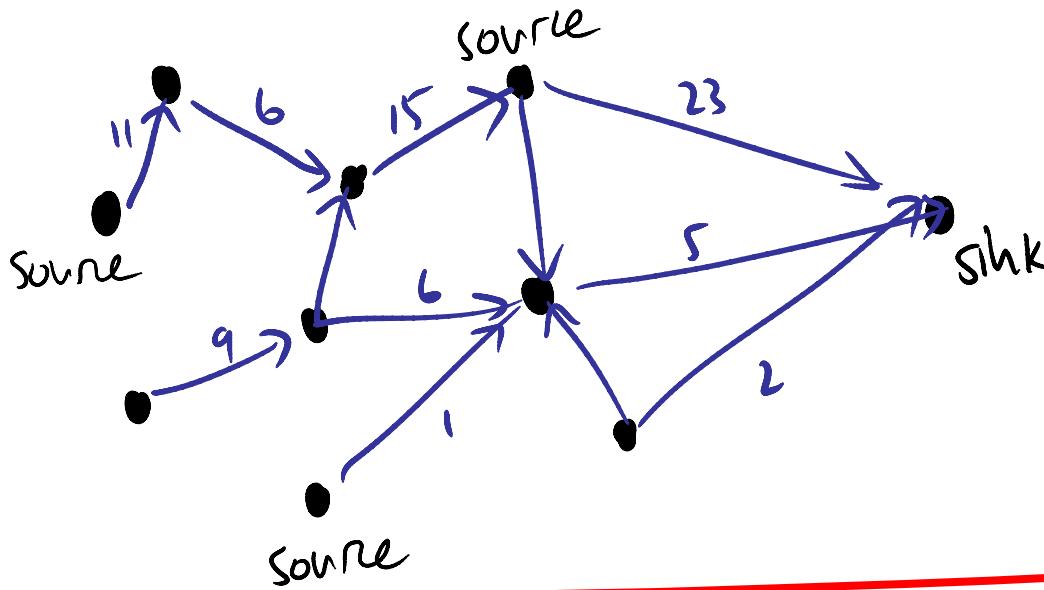


$c(e)$ = cost of edge e
= number of queries skipped by e

$$C_{\text{query}} = \beta \sum_{\text{edges } e} c(e)$$

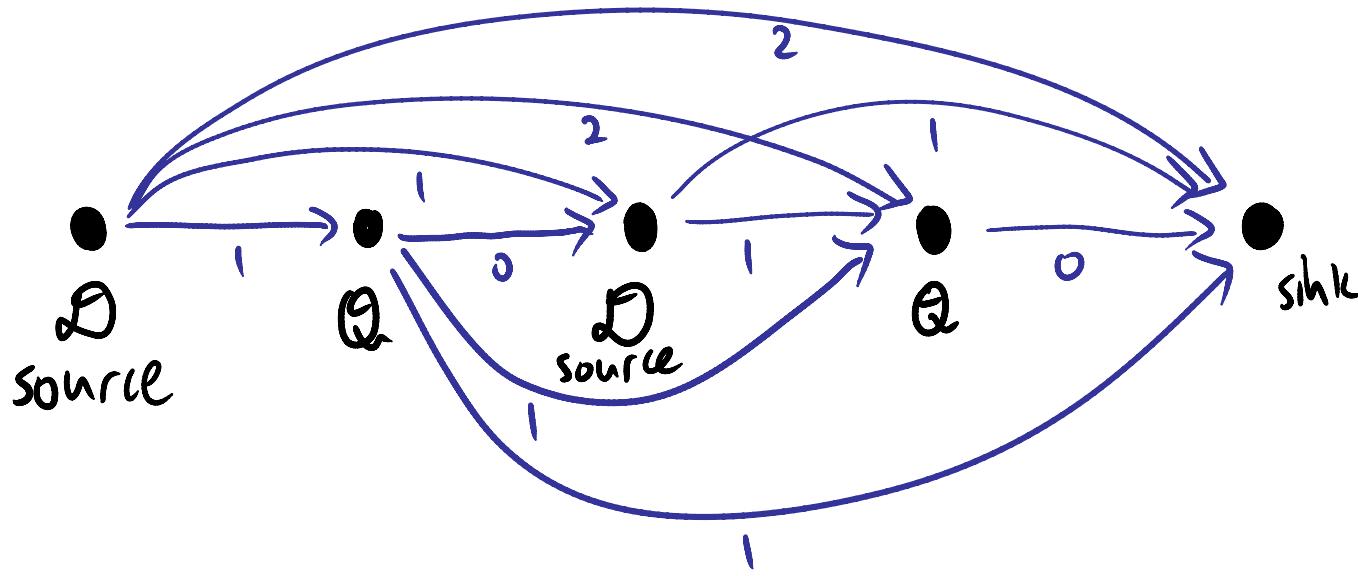
↑
query price

Index merge scheduling is a special case of “directed cost-distance” problem



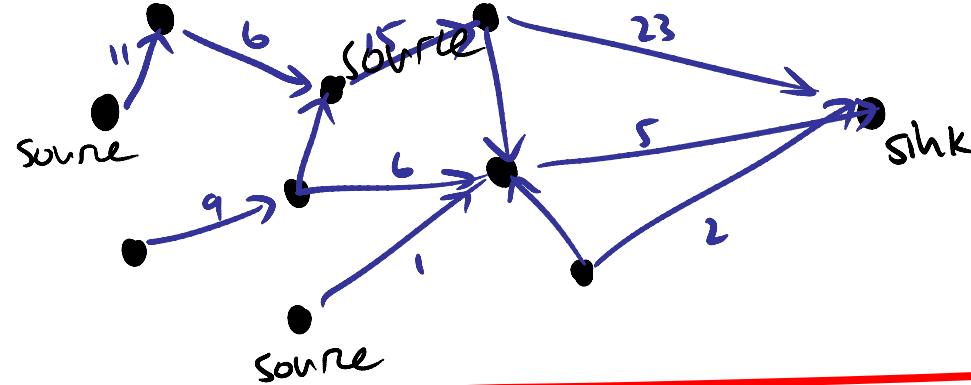
choose a subgraph minimizing $\sum_{edges} \text{edge cost} + \sum_{sources} \text{path length}$

Index merge scheduling is a special case of “directed cost-distance” problem



choose a subgraph minimizing $\sum_{edges} \text{edge cost} + \sum_{sources} \text{path length}$

"Cost-distance problem"



choose a subgraph minimizing

$$\sum_{\text{edges}} \text{edge cost} + \sum_{\text{sources}} \text{path length}$$

- *undirected* case studied by Meyerson-Munagala-Plotkin 2000
- NP-complete (Steiner tree is special case)
- they give an efficient $O(\log(\text{number of sources}))$ approximation
- *directed* case seems much harder
- fortunately, the graph for index merge scheduling has very special structure

the index merge scheduling problem: roadmap

1. single index is insufficient for dynamic content – need multiple indexes, and therefore need occasional index merges
2. scheduling merges is related to the cost-distance problem in network construction
3. imposing geometrically decreasing index sizes gives good performance – $O(n \log n)$

merging as often as possible has quadratic cost



index:
sizes:

	0	0	0	Q	0	0	0	0	sink
	1	2	3		4		5	6	

cumulative

	1	3	6		10		15	21	
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merge cost:

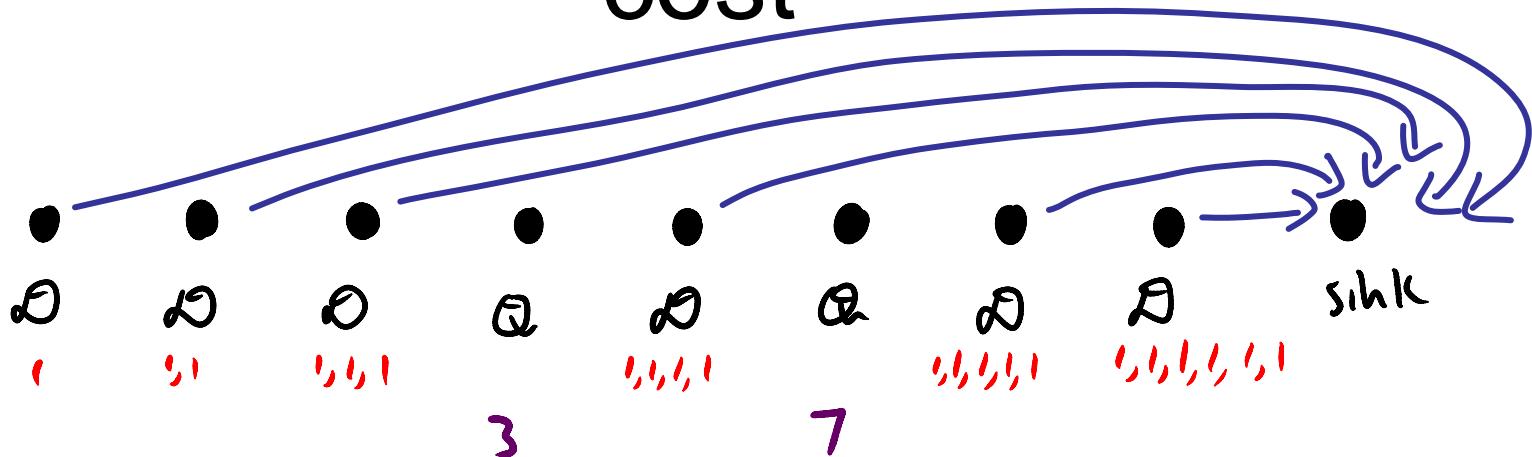
- query cost is minimized (linear in number of queries)
- but merge cost is quadratic in the number of data arrivals:

nth arrival costs αn ,

so after n data arrivals,

$$C_{\text{merge}} = \sum \alpha n(n-1) = O(n^2)$$

never merging also has quadratic cost



index sizes:
cumulative
query cost:

- merge cost is minimized (linear in number of data arrivals)
- but query cost can be quadratic:

e.g. if data and queries alternate $(D, Q, D, Q, \dots, D, Q)$

then

$$C_{\text{query}} = \frac{1}{2} \beta n(n-1) = O(n^2)$$

Maintaining geometrically decreasing index sizes guarantees total I/O cost is $O(T \log T)$

Algorithm:

- Fix $K > 1$ (e.g. $K=2$)

index 1

- Maintain invariant that if index sizes are sorted so that $s_1 > s_2 > \dots > s_n$, then

$$s_i > K s_{i+1} \quad \text{for each } i$$

index 2

- Always merge the r smallest indexes, where r is minimal to keep the invariant

index 3

index 4

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similar alg
used in
Altavista,
SQL Server,
MS Desktop
Search etc

query cost is logarithmic in number of data arrivals

N_D = number of data arrivals

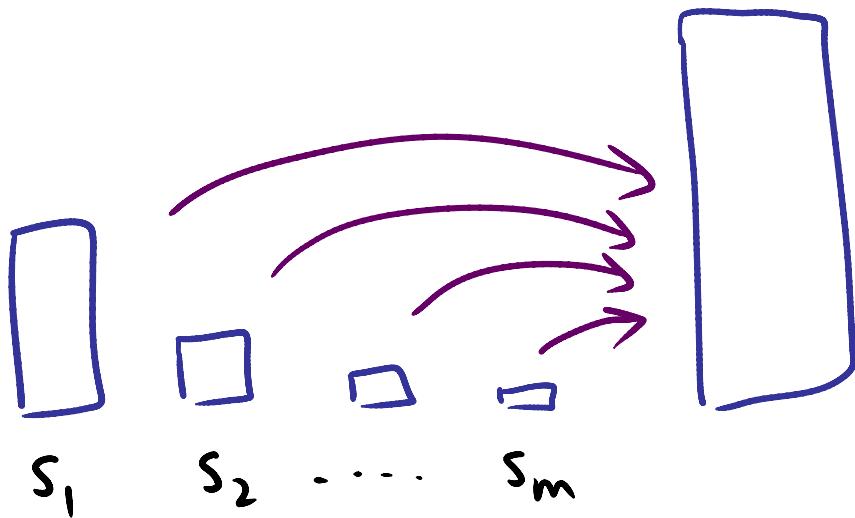
N_Q = number of queries

number of index fibs never exceeds $1 + \log_K N_D$

$$\Rightarrow C_{\text{query}} \leq \beta N_Q (1 + \log_K N_D)$$

merge cost is also logarithmic in number of data arrivals

Lemma If we merge indexes with sizes $s_1 > s_2 > \dots > s_m$, obtaining single new index file of size $s^* = s_1 + s_2 + \dots + s_m$ then each $s_i \leq \frac{K}{K+1} s^*$



Proof: $s_2 + \dots + s_m > \frac{s_1}{K}$
 $\Rightarrow s^* > s_1 + \frac{s_1}{K}$

□

merge cost is also logarithmic in number of data arrivals

Claim Each word-location pair is written out at most $1 + \log_{\frac{K+1}{K}} N_D$ times



merge cost is also logarithmic in
number of data arrivals

Thus,

$$C_{\text{merge}} \leq \alpha N_D \left(1 + \log_{\frac{K+1}{K}} N_D \right)$$

merge cost is also logarithmic in
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Thus,

$$C_{\text{merge}} \leq \alpha N_D \left(1 + \log_{\frac{K+1}{K}} N_D \right)$$

↑ tradeoff in
choice of K

[Recall, $C_{\text{query}} \leq \beta N_Q \left(1 + \log_K N_D \right) \right]$

merge cost is also logarithmic in
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Thus,

$$C_{\text{merge}} \leq \alpha N_D \left(1 + \log_{\frac{k+1}{K}} N_D \right)$$

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[Recall, $C_{\text{query}} \leq \beta N_Q \left(1 + \log_K N_D \right) \right]$

NB For $k=2$, $C_{\text{merge}} \approx \alpha N_D \log_2 N_D$

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2. scheduling merges is related to the cost-distance problem in network construction
3. imposing geometrically decreasing index sizes gives good performance – $O(n \log n)$
4. $O(n \log n)$ is optimal, in general

$O(T \log T)$ is optimal

Consider the input sequence $(D, Q, D, Q, \dots, D, Q)$
 n pairs of D, Q

Claim If n is a power of 2, the optimal merge schedule costs at least

$$\min(\alpha, \beta) \frac{n}{2} \log_2 n$$

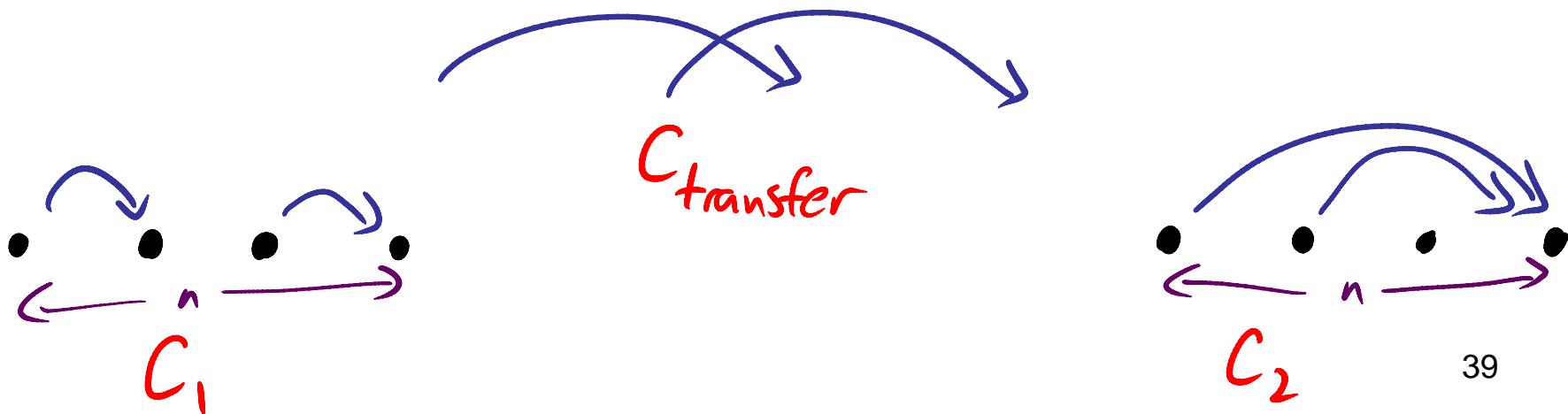
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Proof: Induction on n . Assume for n , prove for $2n$.
Consider optimal strategy for $2n$:



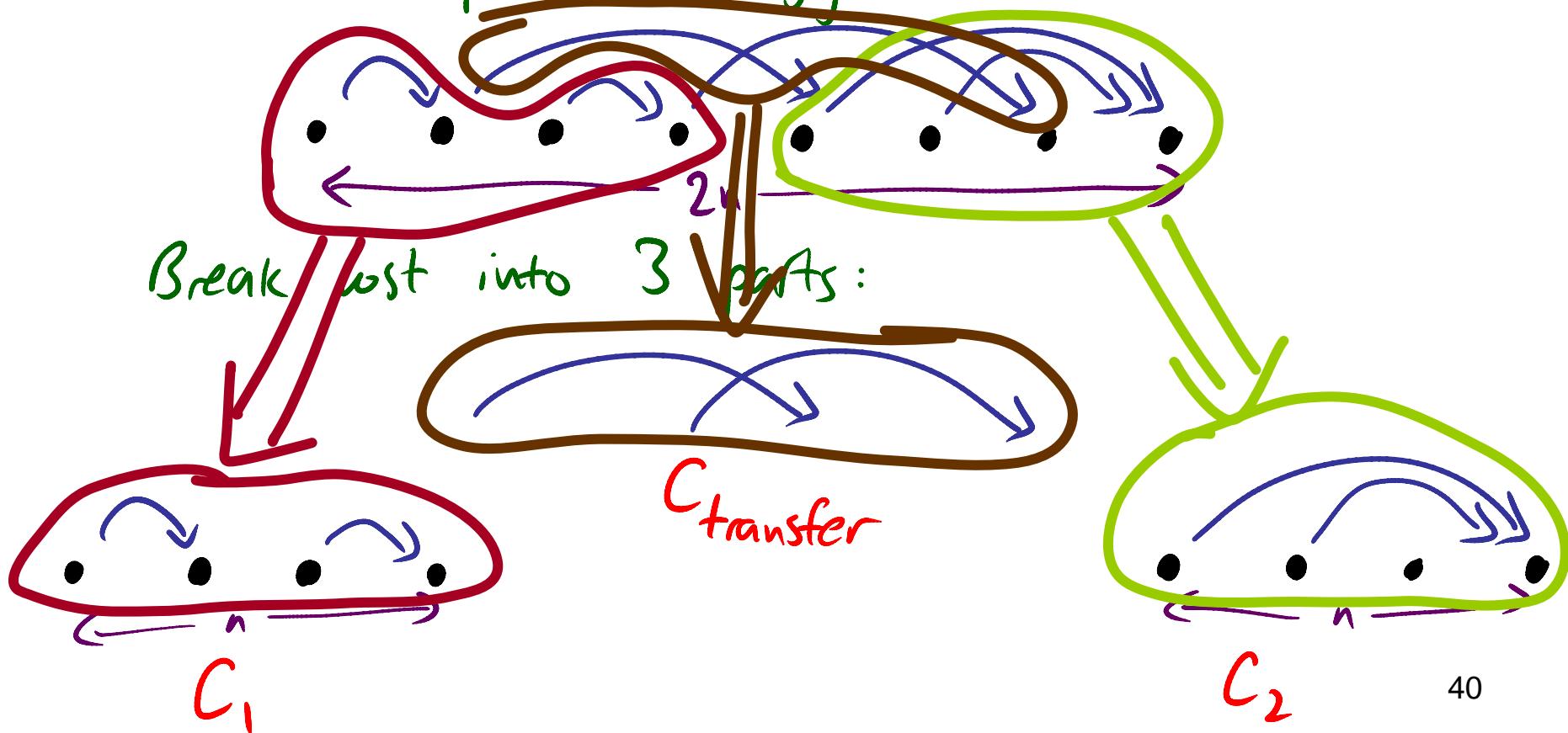
Break cost into 3 parts:



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Claim: C_{transfer} is at least $\min(\alpha, \beta) n$

proof: Either every index file created in 1st half gets rewritten in 2nd half

\Rightarrow additional merging cost of at least αn

or some index file created in 1st half is not rewritten during 2nd half

\Rightarrow additional query cost of at least βn



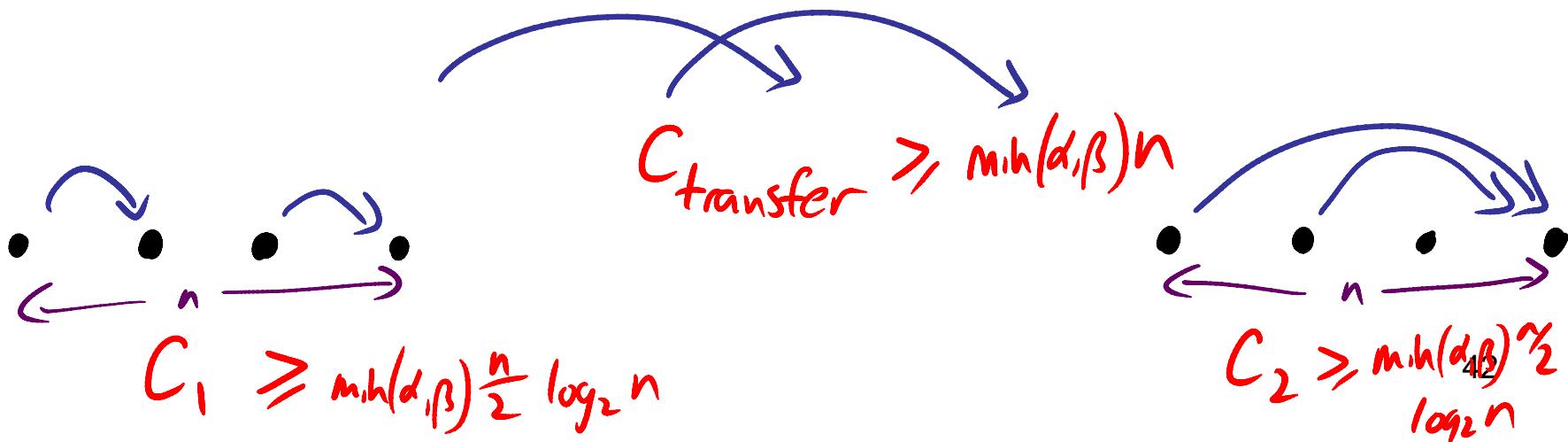
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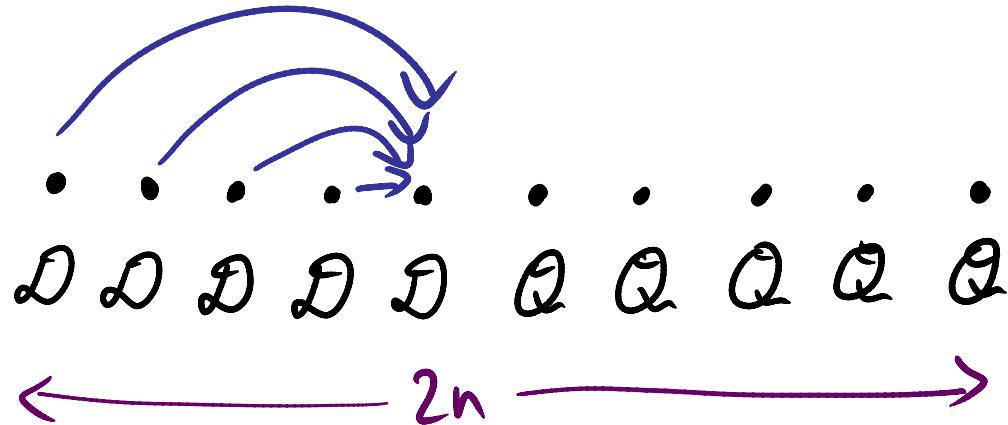
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the index merge scheduling problem: roadmap

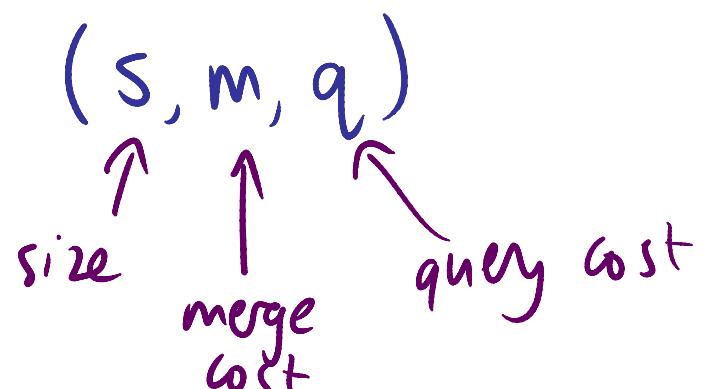
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4. $O(n \log n)$ is optimal, in general
5. cost-balancing approach is more flexible and may be superior – $O(n)$ at times

Some input sequences can be processed in linear cost



$$\text{total cost: } (2\alpha + \beta)n$$

cost-balancing approach is promising

- for every index file, store historic merge + query costs
i.e. index file described by (s, m, q)


size merge cost query cost
- When merging $(s_1, m_1, q_1), (s_2, m_2, q_2), \dots (s_r, m_r, q_r)$,
obtain $(\sum s_i, \sum (m_i + \alpha s_i), \sum q_i)$
- On each query, $q_i \mapsto q_i + \beta$

cost-balancing approach is promising

invariant : $M_i \leq q_i$ [vaguely analogous to ski rental]
"balance"

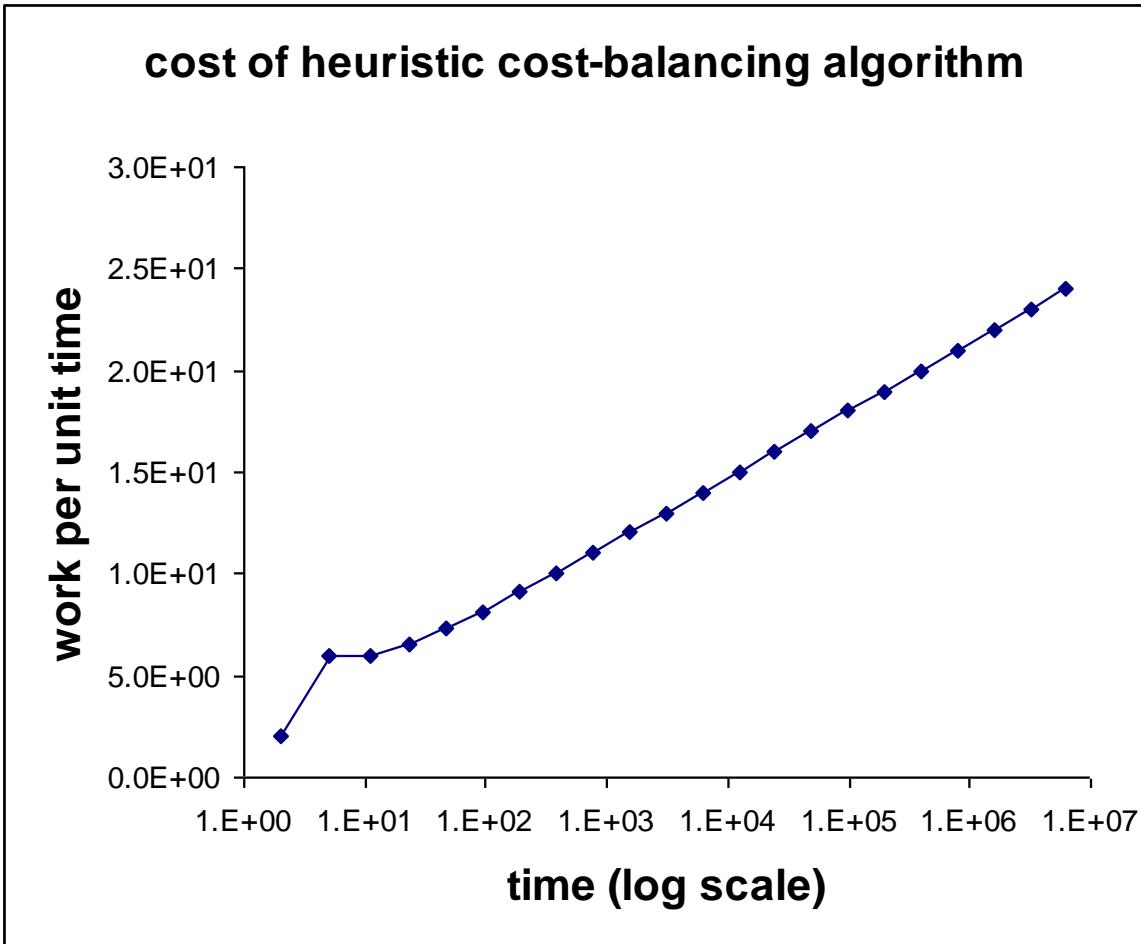
strategy :

- always merge the r smallest index files
- always merge as many as possible without violating balance

cost-balancing approach is promising

- $O(n \log n)$ cost on alternating $DQDQ\dots$ input ?
- $O(n)$ cost on $DDD\dots DQQ\dots Q$ input .
- much more complex/realistic cost modelling is possible

empirical performance of cost-balancing on DQDQ... input is $O(n \log n)$



the index merge scheduling problem: summary

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