

How much information about the future is needed ?

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ski-rental problem

dilemma of a skier:

- rent equipment for 10 EUR per day
- buy equipment for 10c EUR

doesn't know how many days will be skiing

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output $\mathbf{y} = \mathcal{A}(\mathbf{x}) = \langle y_1, y_2, \dots, y_n \rangle$, where $y_i = f(x_1, \dots, x_i)$.

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Theorem [Karp 1992]

optimal worst-case strategy:

rent $c - 1$ days, then buy

competitive ratio:

$$\frac{\text{cost}(\mathcal{A}(\mathbf{x}))}{\text{cost}(\text{OPT}(\mathbf{x}))} \leq \frac{2c - 1}{c} = 2 - \frac{1}{c}$$

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Definition: problem complexity

Best attainable competitive ratio.

other approaches

- **loose competitiveness, ...**: tailored for Paging
- resource augmentation: OPT vs. k -times "more powerful" online
- look-ahead: alg. can see a number of future inputs
- online vs online: Max/Max ratio, relative worst-order ratio, ...
- limited adversary: access graph, statistical, diffuse, ...
- entropy

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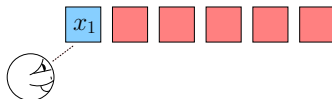
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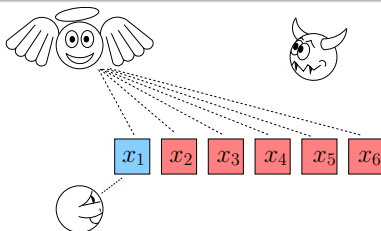
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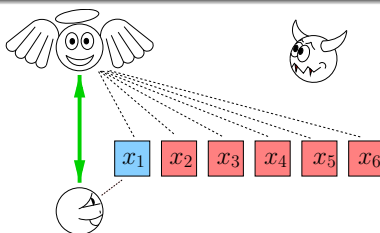
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- algorithm can see part of the input
- oracle sees the whole input
- they can communicate



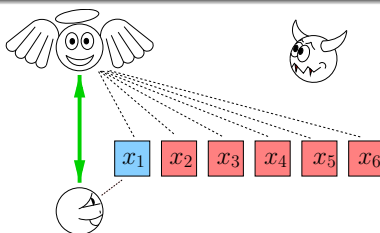
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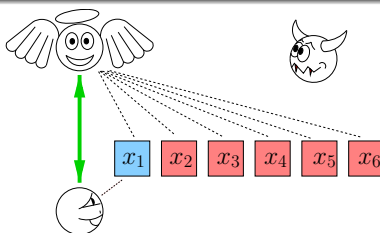
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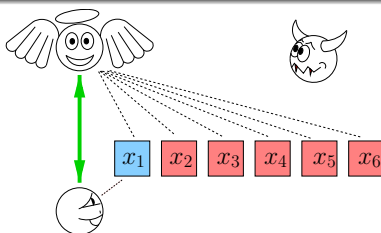
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problem complexity \approx # bits to achieve *OPT*



communication

- **answerer**: algorithm asks, gets an answer
- **helper**: oracle can give spontaneous advice

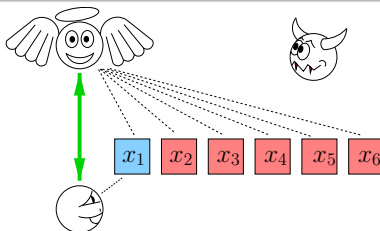


communication

- **answerer**: algorithm asks, gets an answer
- **helper**: oracle can give spontaneous advice

trivial bounds

- encode whole input
- encode output



Definition: helper

input $\mathbf{x} = \langle x_1, x_2, \dots, x_n \rangle$

helper sequence: $\mathcal{O}(\mathbf{x}) = \langle \mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n \rangle$

output: $\mathbf{y} = \langle y_1, y_2, \dots, y_n \rangle$, $y_i = f(x_1, \dots, x_i, \mathbf{a}_1, \dots, \mathbf{a}_i)$

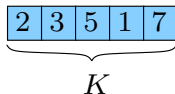
advice (bit) complexity:

$$B_{(\mathcal{A}, \mathcal{O})}^H = \limsup_{n \rightarrow \infty} \max_{|\mathbf{x}|=n} \frac{\sum_{i=1}^n |\mathbf{a}_i|}{n}$$

our results

	competitive ratio	helper	answerer
PAGING	K	$(0.1775, 0.2056)$	$(0.4591, 0.5 + \varepsilon)$
DIFFSERV	≈ 1.281	$\frac{1}{K}$	$\left(\frac{\log K}{2K}, \frac{\log K}{K}\right)$

4 4 3 7

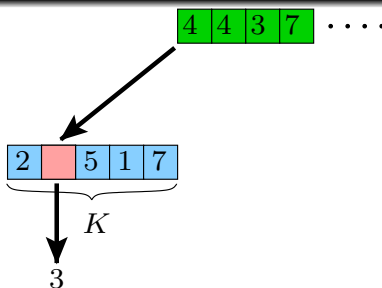


Paging

- input: logical pages $\mathbf{x} = \langle x_1, x_2, \dots, x_n \rangle$, $x_i > 0$
- buffer: physical memory $B = \{b_1, \dots, b_K\}$
- if $x_i \notin B \Rightarrow$ page fault, a victim has to be found

Theorem [Sleator, Tarjan 1985]

Every deterministic algorithm is at least K competitive.



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Replace the farthest-requested page.

online algorithm with helper

1 bit per input request \rightarrow can be optimized

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- a page is **active**, if shall be used by OPT
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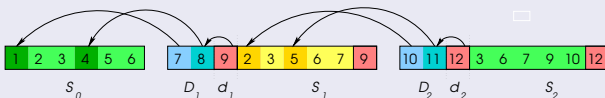
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lower bound



- input: S_0 , blocks of length $3K$

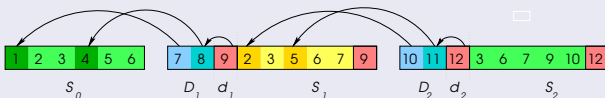
- optimal cannot generate fault in S_j
- distinct inputs have distinct advice sequences
- the number of distinct inputs is

$$Y = \binom{3K-1}{2K-1} = \left[\frac{2}{3} \binom{3K}{K} \right]^j$$

- the number of advice sequences with at most s bits is X :

$$\log X \leq s \left(\log(1 + \alpha) + 1 + \frac{1}{\ln 2} \right) + \frac{1}{2} \left[\log \left(1 + \frac{1}{\alpha} \right) + \log s \right] + c$$

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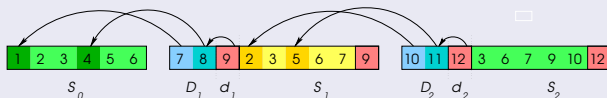
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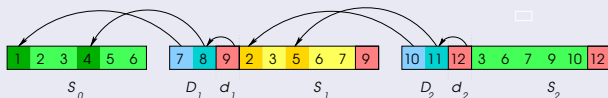
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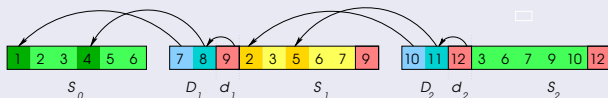
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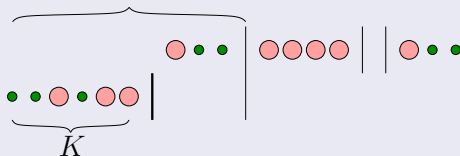
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DiffServ



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skipped

Motivation

- problem complexity \approx amount of **relevant** information
- possible applications – semi-online setting

Open directions

Thank you for your attention

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 - bounded advice
 - trade-off between approximation and information
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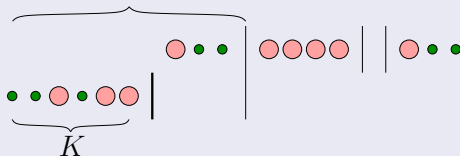
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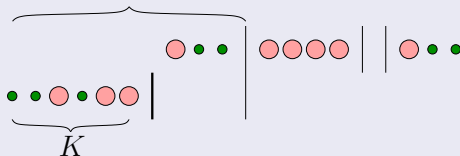
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critical input

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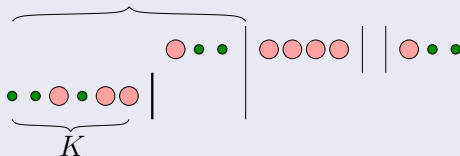
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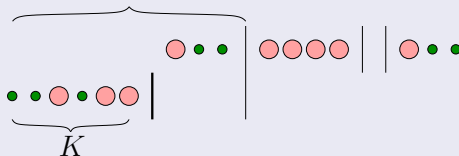
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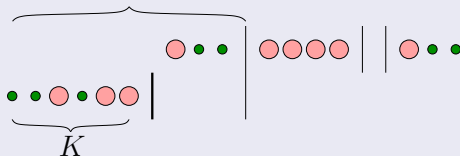
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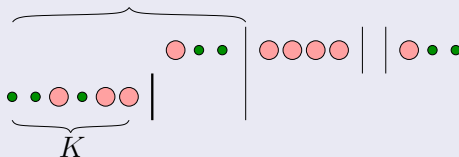
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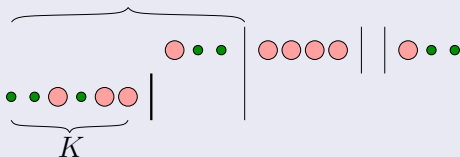
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critical input at most ever $K + 1$ steps $\Rightarrow \frac{1}{K+1}$ bits per input

DiffServ



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