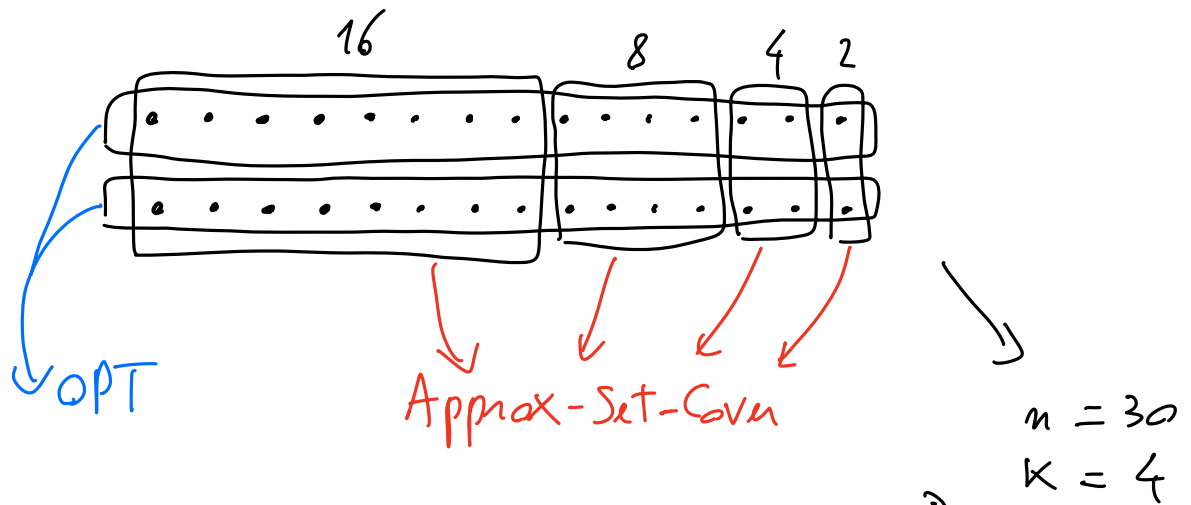


Exercise : show that there is an input  $I = (X, F)$  on which Approx-Set-Cover achieves an approximation ratio of  $\Theta(\log n)$



$X$  has  $n = 2^{(k+1)} - 2$  elements for some  $k \in \mathbb{N}$

$F$  has 1)  $k$  pairwise disjoint sets  $S_1, \dots, S_k$  with sizes  $2, 4, \dots, 2^k$

2) two additional disjoint sets  $T_0, T_1$  each of which contains half of the elements from each  $S_i$

Approx-set-cover  $\rightarrow S_k, S_{k-1}, \dots, S_1$

OPT  $\rightarrow T_0, T_1$

ratio :  $k/2 = \Theta(\log n)$

# Randomized Algorithms

are algorithms that may do random choices ....  
.... but why? Seems paradoxical!

Example 1 : Randomized Quicksort

  $n$  elements

$$T_{QS}(n) = O(n^2)$$

RQS: choose the pivot at random

$$E[T_{RQS}(n)] = O(n \log n)$$

more discussion  
in Further reading

This hides the worst-case inputs  
from the adversary

↳ does no longer know  
algorithm's moves  
in advance

Example 2 : verifying polynomial identities

check whether

$$\begin{array}{ccc} (x+1)(x-2)(x+3)(x-4)(x+5)(x-6) & \stackrel{?}{=} & x^6 - 7x^3 + 25 \\ \parallel & & \parallel \\ H(x) & & G(x) \end{array}$$

obvious algorithm : transform  $H(x)$  in canonical form  $\sum_{i=0}^{d=6} c_i x^i$  and verify whether all the coefficients  $c_i$  of all monomials are equal

$d$  = maximum degree

Complexity :  $O(d^2)$

a faster algorithm:

- choose a random integer  $r$  // new operation
- compute  $H(r)$  //  $O(d)$
- compute  $G(r)$  //  $O(d)$
- if  $H(r) = G(r)$  then return YES
- else return NO

Does it work?

example:  $r=2$

$$H(2) = 0$$

$$G(2) = 33$$

$$\Rightarrow H(x) \neq G(x)$$

what if  $H(r) = G(r)$ ?

$$\text{example: } x^2 + 7x + 1 \stackrel{?}{=} (x+2)^2$$

$$r=2: 19 \neq 16$$

$$r=1: 9 = 9$$

↓  
unlucky  
choice of  $r$

→ alg. returns YES, but it  
is wrong!

If the equation is correct, the algorithm is always correct. Otherwise, the algorithm returns the wrong answer only if  $r$  is a root of the polynomial  $F(x) = G(x) - H(x) = 0$

If  $r \in \{1, 2, \dots, 100d\}$  where  $d$  is the max degree in  $F(x)$ , then

$$\Pr(\text{algorithm fails}) \leq \frac{d}{100d} = \frac{1}{100}$$

small, but still not  
satisfactory

How to reduce the probability of error?

- run the algorithm 10 times
- if YES in all the 10 times then return YES
- else return NO

Now

$$\Pr(\text{algorithm fails}) \leq \left(\frac{1}{100}\right)^{10} = 10^{-20} < 2^{-64}$$

$2^{-64}$  is comparable to the probability of a hardware error in your computer caused by cosmic radiation (quoting D. Knuth). So, this alg. is correct for all practical purposes.

## Classification of randomized algorithms

1) rand. alg. that never fail  $\rightarrow$  "LAS VEGAS" alg.  
e.g. randomized Quicksort

$$\forall i \in I, A_R(i) = s \xrightarrow{\text{solution}} \text{s.t. } (i, s) \in \Pi$$

obs.:  $s$  may not be the same  $\forall i$

randomness come into play in the analysis of the complexity

$\forall n$ ,  $T(n)$  is a random variable, of which we usually study  $E[T(n)]$  or

$$\Pr(T(n) > c \cdot f(n))$$

$$\hookrightarrow \leq \frac{1}{n^k} \rightarrow T(n) = O(f(n)) \text{ "with high probability"}$$

space of probabilities = random choices made by the algorithm

(Do not confuse this with the probabilistic analysis of deterministic algorithm, where the space of probabilities = distribution of the inputs)

2) rand. alg. that may fail  $\rightarrow$  "MONTE CARLO" alg.  
e.g. verifying polynomial identities

$i \in I$  it's possible that  $A_R(i) = s$  s.t.

$$(i, s) \notin \Pi$$

We study  $Pr((i, s) \notin \Pi)$  as a function of  $n = |I| \rightarrow$  family of random variables

moreover, even  $T(n)$  may be a random variable

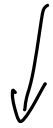
for decision problems, these alg. can be divided into

- one-sided : may fail only on one answer
- two-sided : may fail in both answers

We'll see 1 LAS VEGAS and 1 MONTE CARLO



Randomized Quicksort



Karger's alg. for  
minimum cut