Randomized Algorithms II

## Randomized algorithms

- Last weeks
	- Contention resolution
	- Global minimum cut
- Today
	- Expectation of random variables
		- Guessing cards
	- Selection
	- Quicksort





# Random Variables and Expectation

### Random variables

- A random variable is an entity that can assume different values.
- The values are selected "randomly"; i.e., the process is governed by a probability distribution.
- Examples: Let X be the random variable "number shown by dice".
	- $\cdot$  X can take the values 1, 2, 3, 4, 5, 6.
	- $\cdot$  If it is a fair dice then the probability that  $X = 1$  is 1/6:
		- Pr[X=1] =  $1/6$ .
		- Pr[X=2] =  $1/6$ .

• …

#### Expected values

- Let X be a random variable with values in  $\{x_1,...x_n\}$ , where  $x_i$  are numbers.
- The expected value (expectation) of X is defined as

$$
E[X] = \sum_{j=1}^{n} x_j \cdot \Pr[X = x_j]
$$

- The expectation is the theoretical average.
- Example:
	- $\cdot$  X = random variable "number shown by dice"

$$
E[X] = \sum_{j=1}^{6} j \cdot \Pr[X=j] = (1+2+3+4+5+6) \cdot \frac{1}{6} = 3.5
$$

### Waiting for a first succes

- Coin flips. Coin is heads with probability  $p$  and tails with probability  $1 p$ . How many independent flips X until first heads?
	- Probability of  $X = j$ ? (first succes is in round  $j$ )

$$
\Pr[X=j] = (1-p)^{j-1} \cdot p
$$

• Expected value of  $X$ :

$$
E[X] = \sum_{j=1}^{\infty} j \cdot \Pr[X=j] = \sum_{j=1}^{\infty} j \cdot (1-p)^{j-1} \cdot p = \frac{p}{1-p} \sum_{j=1}^{\infty} j \cdot (1-p)^{j}
$$

$$
= \frac{p}{1-p} \cdot \frac{1-p}{p^2} = \frac{1}{p}
$$

$$
\sum_{k=0}^{\infty} k \cdot x^k = \frac{x}{(1-x)^2} \quad \text{for } |x| < 1.
$$

#### Properties of expectation

- If we repeatedly perform independent trials of an experiment, each of which succeeds with probability  $p>0,$  then the expected number of trials we need to perform until the first succes is . 1/*p*
- $\cdot$  If  $X$  is a 0/1 random variable, then  $E[X] = \Pr[X = 1].$
- Linearity of expectation: For two random variables X and Y we have

 $E[X + Y] = E[X] + E[Y]$ 

# Guessing cards

- Game. Shuffle a deck of  $n$  cards; turn them over one at a time; try to guess each card.
- Memoryless guessing. Can't remember what's been turned over already. Guess a card from full deck uniformly at random.
- Claim. *The expected number of correct guesses is 1*.
	- $\cdot$   $X_i = 1$  if  $i^{th}$  guess correct and zero otherwise.
	- $\cdot$   $X =$  the correct number of guesses  $X_1 + \ldots + X_n$ .
	- $E[X_i] = Pr[X_i = 1] = 1/n$ .
	- $E[X] = E[X_1 + \dots + X_n] = E[X_1] + \dots + E[X_n] = 1/n + \dots + 1/n = 1.$





### Guessing cards

Game. Shuffle a deck of n cards; turn them over one at a time; try to guess each card.

 $\ln n < H(n) < \ln n + 1$ 

- Guessing with memory. Guess a card uniformly at random from cards not yet seen.
- Claim. *The expected number of correct guesses is*  $\Theta(\log n)$ .
	- $\cdot$   $X_i = 1$  if  $i^{th}$  guess correct and zero otherwise.
	- $\cdot$   $X =$  the correct number of guesses  $X_1 + \ldots + X_n$ .
	- $E[X_i] = Pr[X_i = 1] = 1/(n i + 1).$
	- $E[X] = E[X_1] + \dots + E[X_n] = 1/n + \dots + 1/2 + 1/1 = H_n$ .

### Coupon collector

- Coupon collector. Each box of cereal contains a coupon. There are *n* different types of coupons. Assuming all boxes are equally likely to contain each coupon, how many boxes before you have at least 1 coupon of each type?
- Claim. *The expected number of steps is*  $\Theta(n\log n)$ .
	- Phase  $j$  = time between  $j$  and  $j + 1$  distinct coupons.
	- $\cdot$   $X_j$  = number of steps you spend in phase  $j$ .
	- $\cdot$   $X =$  number of steps in total =  $X_0 + X_1 + \dots + X_{n-1}$ .
	- $E[X_j] = n/(n-j)$ .
	- The expected number of steps:

$$
E[X] = E[\sum_{j=0}^{n-1} X_j] = \sum_{j=0}^{n-1} E[X_j] = \sum_{j=0}^{n-1} n/(n-j) = n \cdot \sum_{i=1}^{n} 1/i = n \cdot H_n.
$$

Median/Select

#### **Select**

- Given *n* numbers  $S = \{a_1, ..., a_n\}$ .
- Median: number that is in the middle position if in sorted order.
- $\cdot$  Select( $S$ , $k$ ): Return the  $k$ th smallest number in  $S.$ 
	- $\cdot$  Min(S) = Select(S, 1), Max(S)= Select(S, n), Median = Select(S, n/2).
- Assume the numbers are distinct.

```
Select(S, k) 
 Choose a pivot s ∈ S uniformly at random. 
 For each element e in S: 
if e < s put e in S' 
if e > s put e in S'' 
 if |S'| = k-1 then return s 
if |S'| \geq k then call Select(S', k)
if |S'| < k then call Select(S'', k - |S'| - 1)
```
#### Select: Running time

```
Select(S, k) 
 Choose a pivot s ∈ S uniformly at random. 
 For each element e in S: 
if e < s put e in S' 
if e > s put e in S'' 
 if |S'| = k-1 then return s 
if |S'| \ge k then call Select(S', k)
if |S'| < k then call Select(S'', k - |S'| - 1)
```
- $\cdot$  Worst case running time:  $T(n) = cn + c(n − 1) + c(n − 2) + ... + c = Θ(n^2)$
- $\cdot$  If there is at least an  $\varepsilon$  fraction of elements both larger and smaller than s:

$$
T(n) = cn + (1 - \varepsilon)cn + (1 - \varepsilon)^2 cn + \dots
$$
  
= (1 + (1 - \varepsilon) + (1 - \varepsilon)^2 + \dots)cn  

$$
\le cn/\varepsilon
$$

 $\cdot$  Intuition: A fairly large fraction of elements are "well-centered"  $\Rightarrow$  random pivot likely to be good.

## Select: Analysis

• Central element:  $\geq 1/4$  of the elements in current S are smaller and  $\geq 1/4$  are larger.

S

- If pivot central: size of set shrinks by at least a factor 3/4.
- At least half the elements are central  $\Rightarrow$  Pr[s is central] = 1/2. ⇒
- Phase j: Size of set at most  $(3/4)^j n$  and at least  $(3/4)^{j+1} n$ .
	- Pivot central  $\Rightarrow$  current phase ends.
	- $\cdot$  Expected number of iterations before a central pivot is found = 2.
- $\boldsymbol{X}$  = number of steps taken by algorithm.  $X_j$  = number of steps in phase  $j$ .
- Then  $X = X_1 + X_2 + \cdots$
- $E[X_j] = 2cn(3/4)^j$
- Expected running time:

$$
E[X] = E\left[\sum_{j} X_{j}\right] = \sum_{j} E[X_{j}] = \sum_{j} 2cn\left(\frac{3}{4}\right)^{j} = 2cn\sum_{j} \left(\frac{3}{4}\right)^{j} \leq 8cn
$$

**Quicksort** 

## **Quicksort**

- Given *n* numbers  $S = \{a_1, ..., a_n\}$ .
- Assume the numbers are distinct.

```
Quicksort(S) 
 if |S| ≤ 1 return S 
 else 
  Choose a pivot s ∈ S uniformly at random. 
  For each element e in S 
    if e < s put e in S' 
    if e > s put e in S'' 
L = Quicksort(S') 
R = Quicksort(S'') 
return the sorted list L◦s◦R
```
## Quicksort: Analysis

- Worst case:  $\Omega(n^2)$  comparisons.
- Best case: *O*(*n* log *n*)
- Enumerate elements such that  $a_1 \le a_2 \le \cdots \le a_n$ .
- Indicator random variable for all pairs  $i < j$ :

$$
X_{ij} = \begin{cases} 1 & \text{if } a_i \text{ and } a_j \text{ are compared by the algorithm} \\ 0 & \text{otherwise} \end{cases}
$$

 $\cdot$   $\,X$  total number of comparisons:

$$
X = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} X_{ij}
$$

• Expected number of comparisons:

$$
E[X] = E[\sum_{i=1}^{n-1} \sum_{j=i+1}^{n} X_{ij}] = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} E[X_{ij}]
$$

#### Quicksort: Analysis

- Compute expected number of comparisons.
- Since  $X_{ij}$  is an indicator variable:  $E[X_{ij}] \; = \; \Pr[X_{ij} = 1].$
- $a_i$  and  $a_j$  compared  $\Leftrightarrow a_i$  or  $a_j$  is the first pivot element chosen from  $Z_{ij} = \{a_i, ..., a_j\}$
- Pivot chosen independently uniformly at random  $\Rightarrow$

all elements from  $Z_{ij}$  equally likely to be chosen as first pivot from this set.

- We have  $Pr[X_{ij} = 1] = 2/(j i + 1)$ .
- Thus

$$
E[X] = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} E[X_{ij}] = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \Pr[X_{ij} = 1] = \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \frac{2}{j-i+1}
$$
  
= 
$$
\sum_{i=1}^{n-1} \sum_{k=2}^{n-i+1} \frac{2}{k} \le \sum_{i=1}^{n-1} \sum_{k=1}^{n} \frac{2}{k} = 2 \sum_{i=1}^{n-1} H_n = 2n \cdot H_n \le O(n \log n)
$$