

## Randomized Algorithms II

## Randomized algorithms

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



## Random Variables and Expectation

## Random variables

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- 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”.
  - $X$  can take the values 1, 2, 3, 4, 5, 6.
  - 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  $\{x_1, \dots, x_n\}$ , 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:

- $X$  = 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  $X$ :

$$\begin{aligned} E[X] &= \sum_{j=1}^{\infty} j \cdot \Pr[X = j] = \sum_{j=1}^{\infty} j \cdot (1 - p)^{j-1} \cdot p = \frac{p}{1 - p} \sum_{j=1}^{\infty} j \cdot (1 - p)^j \\ &= \frac{p}{1 - p} \cdot \frac{1 - p}{p^2} = \frac{1}{p} \end{aligned}$$

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

## 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  $1/p$ .

- 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.

- $X_i = 1$  if  $i^{\text{th}}$  guess correct and zero otherwise.

- $X$  = the correct number of guesses =  $X_1 + \dots + X_n$ .

- $E[X_i] = \Pr[X_i = 1] = 1/n$ .

- $E[X] = E[X_1 + \dots + X_n] = E[X_1] + \dots + E[X_n] = 1/n + \dots + 1/n = 1$ .



## Guessing cards

- **Game.** Shuffle a deck of  $n$  cards; turn them over one at a time; try to guess each card.
- **Guessing with memory.** Guess a card uniformly at random from cards not yet seen.
- **Claim.** The expected number of correct guesses is  $\Theta(\log n)$ .
  - $X_i = 1$  if  $i^{\text{th}}$  guess correct and zero otherwise.
  - $X =$  the correct number of guesses  $= X_1 + \dots + X_n$ .
  - $E[X_i] = \Pr[X_i = 1] = 1/(n - i + 1)$ .
  - $E[X] = E[X_1] + \dots + E[X_n] = 1/n + \dots + 1/2 + 1/1 = H_n$ .

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

## 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.
  - $X_j =$  number of steps you spend in phase  $j$ .
  - $X =$  number of steps in total  $= X_0 + X_1 + \dots + X_{n-1}$ .
  - $E[X_j] = n/(n - j)$ .
  - The expected number of steps:

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

## Median/Select

## Select

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

**Select(S, k)**

Choose a pivot  $s \in S$  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'| ≥ 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  $s \in S$  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( $S''$ ,  $k - |S'| - 1$ )
```

• Worst case running time:  $T(n) = cn + c(n-1) + c(n-2) + \dots + c = \Theta(n^2)$

• If there is at least an  $\epsilon$  fraction of elements both larger and smaller than  $s$ :

$$\begin{aligned} T(n) &= cn + (1-\epsilon)cn + (1-\epsilon)^2cn + \dots \\ &= (1 + (1-\epsilon) + (1-\epsilon)^2 + \dots)cn \\ &\leq cn/\epsilon \end{aligned}$$

• **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.



• If **pivot central**: size of set shrinks by at least a factor  $3/4$ .

• **At least half** the elements are central  $\Rightarrow \Pr[s \text{ is central}] = 1/2$ .

• **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.

• Expected number of iterations before a central pivot is found = 2.

•  $X$  = number of steps taken by algorithm.  $X_j$  = number of steps in phase  $j$ .

• Then  $X = X_1 + X_2 + \dots$

•  $E[X_j] = 2cn(3/4)^j$

• **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  $S = \{a_1, \dots, a_n\}$ .

• Assume the numbers are distinct.

```
Quicksort(S)  
  
if  $|S| \leq 1$  return  $S$   
  
else  
  Choose a pivot  $s \in S$  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  $L \circ s \circ R$ 
```

## Quicksort: Analysis

• Worst case:  $\Omega(n^2)$  comparisons.

• Best case:  $O(n \log n)$

• Enumerate elements such that  $a_1 \leq a_2 \leq \dots \leq a_n$ .

• Indicator random variable for all pairs  $i < j$ :

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

•  $X$  total number of comparisons:

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

• Expected number of comparisons:

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

## 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  $Z_{ij} = \{a_i, \dots, a_j\}$

• 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

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