

Randomization

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Randomized algorithms

- Make random choices (coin flips, random numbers ..)
- Different random choices are assumed independent
- Outcome of algorithm and running time depends on random choices (besides input)
- **Correctness:** Show termination and correct answer for all random choices

Basic probability concepts

- **Sample space:** Set of all possibilities (=sample points).
- We'll deal with finite, or countable sample spaces.
- **Examples:** Flip 100 coins: 2^{100} sample points= $\{H,T\}^{100}$
 - Inputs of size $n=100$ numbers in a range $[0,10^{20}]$ (eg. sorting)
 - Set of possible random choices w of a randomized algorithm when run on a specific input
- Probability of each sample point: $\text{sum}=1$
 - Ex. 100 independent unbiased coin flips:
probability of each sample point (outcome)= 2^{-100}

Basic probability concepts

- **Event** A = subset of sample space
- $\Pr(A) = \text{sum} \{ \Pr(s) \mid s \in A \}$
 - Ex. A = 3 heads in 100 coin flips where coin bias = p:

$$\Pr(A) = \binom{100}{3} p^3 (1-p)^{97}$$

- Independent events A, B:
 $\Pr(A \text{ and } B) = \Pr(A)\Pr(B)$

Random variables , Expectation

- **Random variable** X : Maps Sample space S to R
 - Ex: # heads : binomial distribution
 - Time $t(I)$ of a deterministic algorithm for input I of size n
 - $t(I,w)$ of a randomized algorithm for particular input I , random choices w
- If X is a discrete random variable (eg. #Hs in coin flip, or time $t(I,w)$ for particular input I , random choices w) , then $E[X] = \sum_x x \text{Prob}[X=x]$
- **Average case (expected) time complexity of deterministic algorithm for an input probability distribution** (probability distribution on inputs of size n for every n):

$$\overline{T}(n) = E_{|I|=n} t(I)$$

Time complexity of randomized algorithms

- **Running time:** Depends on input I and random choices w : time $t(I, w)$
- **Expected running time for an input I :**
expected time w.r.t. random choices w : $\bar{t}(I) = E_w t(I, w)$
- **Expected time complexity of the algorithm**

Two versions:

- **Worst-case expected time** $T(n) = \max_{|I|=n} E_w t(I, w)$
(worst-case expected time over inputs of size n)
- **Average-case expected time** $\bar{T}(n) = E_{|I|=n} E_w t(I, w)$
assumes a probability distribution on inputs of size n , expected time also w.r.t. inputs

Expectation – Indicator r.v.

- Linearity of expectations $E[X+Y] = E[X]+E[Y]$
- Often it is hard to compute the probabilities $\text{Prob}[X=x]$ and do the summation in the defn of $E[X]$
- If can decompose X as sum $Z_1 + \dots + Z_m$ of simpler variables then $E[X] = E[Z_1] + \dots + E[Z_m]$
- **Indicator random variable $I(A)$:** indicates if event A happens: 1 if it happens, 0 if it does not
- $E[I(A)] = 1 \text{ Prob}(A) + 0 \text{ Prob}(\text{not } A) = \text{Prob}(A)$
- **Example: n coin flips with bias p , $X = \#H$'s**

$$E[X] = \sum_{i=1}^n i \binom{n}{i} p^i (1-p)^{n-i} = np$$

- $Z_i = I(\text{i-th flip} = H)$. $E(Z_i) = p$
- $X = Z_1 + \dots + Z_n \Rightarrow E[X] = np$

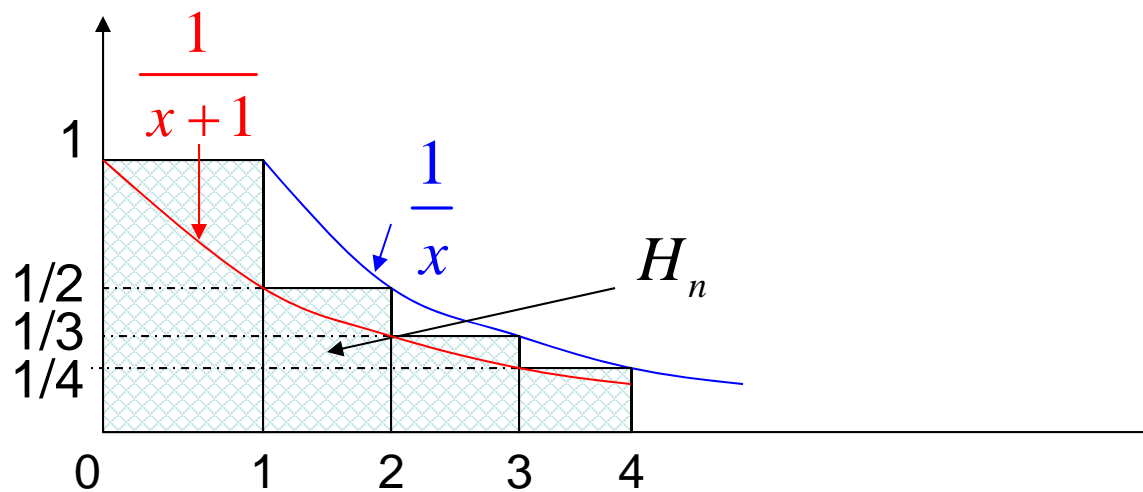
Example: Hiring problem

- Interview n candidates for assistant in order, hire candidate i if better than current asst and fire current asst.
- **Sample space**: orderings of candidates (permutations)
- **Probability Distribution**: uniform: all same probability $=1/n!$
- **Random variable X** : #times hire new asst.
- Indicator variable $Z_i = I(\text{hire } i\text{-th candidate})$
- $E(Z_i) = \text{Prob. that the best among candidates } \{1, \dots, i\} \text{ is } i = 1/i$

$$E[X] = \sum_{i=1}^n \frac{1}{i} = H_n \approx \ln n \quad \text{Harmonic series}$$

Harmonic series

$$E[X] = \sum_{i=1}^n \frac{1}{i} = H_n \approx \ln n$$



$$H_n > \int_0^n \frac{1}{x+1} dx = \ln(n+1) > \ln n$$

$$H_n < 1 + \int_1^n \frac{1}{x} dx = 1 + \ln n$$

Birthday Paradox

- How many people do you need until you expect to find two people with the same birthday?
- Answer: \ll number $n=365$ of days - Grows as $\Theta(\sqrt{n})$, assuming birthdays are independent uniformly random.
- 23 people \Rightarrow at least $\frac{1}{2}$ probability of a birthday coincidence
- Proof: If k people, then $\text{prob}(\text{no coincidence}) =$

$$\frac{n(n-1)\cdots(n-k+1)}{n^n} = 1\left(1-\frac{1}{n}\right)\cdots\left(1-\frac{k-1}{n}\right)$$

Since $1+x \leq e^x$, probability is $\leq e^{-(1+2+\cdots+k-1)/n} = e^{-k(k-1)/2n}$

which is $\leq \frac{1}{2}$ if $k \geq 1 + \sqrt{1 + (8 \ln 2)n} / 2$

For $n=365$, $k \geq 23$ suffices.

Birthday paradox via indicator r.v.

- $X_{ij} = I(\text{persons } i \text{ and } j \text{ have the same birthday})$
- $E[X_{ij}] = \text{Prob}(X_{ij}) = n/n^2 = 1/n$
- $E(\text{\#pairs with same birthday}) = E(\sum X_{ij}) =$
 $= \sum E(X_{ij}) = k(k-1)/2n$

\Rightarrow If $k \geq \sqrt{2n} + 1$ then $E(\text{\#pairs with same birthday}) \geq 1$

For $n=365$, $k \geq 28$ suffices