

# CS5234 Cheat Sheet Gabriel Yeo

## Inequalities

Markov:	$P(X \geq \alpha) \leq \frac{E(X)}{\alpha}$
Chebyshev:	$P( X - \mu  \geq k) \leq \frac{Var(X)}{k^2}$
Chernoff:	$P(X \leq (1 - \delta)\mu) \leq e^{-\frac{\delta^2 \mu}{2}} \leq e^{-\frac{\delta^2 \mu}{3}}$ $P(X \geq (1 + \delta)\mu) \leq e^{-\frac{\delta^2 \mu}{2 + \delta}} \leq e^{-\frac{\delta^2 \mu}{3}}$ $X = X_1 + \dots + X_n$ , $X_i \in \{0, 1\}$ , (ind. Bernoulli)
Chernoff:	$P( X - \mu  \geq \delta\mu) \leq 2e^{-\frac{\delta^2 \mu}{3}}$
Hoeffding:	$P( X - \mu  \geq t) \leq 2e^{-\frac{t^2}{n}}$ $X = X_1 + \dots + X_n$ , $X_i \in [0, 1]$
Hoeffding:	$P( \bar{X} - E[\bar{X}]  \geq t) \leq 2e^{-2nt^2}$
Hoeffding (gen):	$P( \bar{X} - E[\bar{X}]  \geq t) \leq 2e^{-\frac{2t^2}{\sum (a_i - b_i)^2}}$ $X_i \in [a_i, b_i]$

## Facts

Taylor expansion:	$e^x = 1 + x + \frac{x^2}{2!} + \dots = \sum_{n=0}^{\infty} \frac{x^n}{n!}$
For $0 < x \leq 1$ :	$1 - x \leq e^{-x} \leq 1 - \frac{x}{2}$
For $0 < x \leq 1$ :	$\frac{1}{e^2} \leq (1 - x)^{1/x} \leq \frac{1}{e}$
Approx $\binom{a}{b}$ :	$\left(\frac{a}{b}\right)^b \leq \binom{a}{b} \leq \left(\frac{ea}{b}\right)^b$
For $0 < x < 1$ :	$\sum_{i=0}^{\infty} a^i = \frac{1}{1-a}$
Natural log:	$\ln(n-1) \leq \sum_{i=1}^n \frac{1}{i} \leq \ln(n) + 1$
Powers of 2:	$\sum_{i=0}^n 2^i = 2^{n+1} - 1$
1/e:	= 0.36787944117
1/e <sup>2</sup> :	= 0.13533528323

## Probability

Expectation:	$E[X] = \sum_{v \in D} v \cdot Pr[X = v]$
Variance:	$Var[X] = E[(X - E[X])^2]$ $= E[X^2] - E[X]^2$
More variance:	$Var[aX] = a^2 Var[X]$
Linearity of expectation:	$E[\sum X_i] = \sum E[X_i]$
If independent $X_i$ :	$Var[\sum X_i] = \sum Var[X_i]$
Conditional:	$P[X \text{ and } Y] = P[X Y] \cdot P[Y]$
Independence:	$P[X \text{ and } Y] = P[X] \cdot P[Y]$
Union bound:	$P[\bigcup E_i] \leq \sum P[E_i]$

## Complexities

Algorithm	Classical	Approx
BFS	$O(n + m)$	
Connected?	$O(n + m)$	$O(\frac{1}{\epsilon^2 d})$
# CCs	$O(n + m)$	$O(\frac{d}{\epsilon^3})$
MST weight	$O(m \log n)$	$O(\frac{dW^4 \log W}{\epsilon^3})$
Prim's	$O((m + n) \cdot \log n)$	
Prim's (Fib heap)	$O(m + n \log n)$	
Kruskal's (MST)	$O(m \log n)$	
Dijkstra	$O((m + n) \cdot \log n)$	
Dijkstra (Fib heap)	$O(m + n \log n)$	

## Lecture 1 All 0s

Give an array  $A$  with  $n$  elements,  $A[i] \in \{0, 1\}$ :  
 (1) If array has  $\geq \epsilon n$  1's, return False with probability at least  $1 - \delta$ :  
 Assume  $\geq \epsilon n$  1's, then for sample  $i$ :  
 $Pr[A[i] = 1] \geq \epsilon n / n \geq \epsilon$ .

$$\begin{aligned}
 Pr(\text{all samples are 0}) &\leq (1 - \epsilon)^s \\
 &\leq (1 - \epsilon)^{\frac{\ln(1/\delta)}{\epsilon}} \\
 &\leq e^{-\ln(1/\delta)} \\
 &\leq \delta
 \end{aligned}$$

$$\text{Fix } s = \frac{\ln(1/\delta)}{\epsilon}$$

## Lecture 1 Number of 1s

Give an array  $A$  with  $n$  elements,  $A[i] \in \{0, 1\}$ :  
 (1) Find number of 1's  $\pm \epsilon$  with probability at least  $1 - \delta$ .  
 Let  $Y_i$  be  $s$  independent samples in  $\{0, 1\}$ .  
 Output =  $Z = 1/s \sum Y_i$   
 Probability of failure:

$$\begin{aligned}
 Pr(|Z - E[Z]| \geq \epsilon) &\leq 2e^{-2s\epsilon^2} \\
 &\leq 2e^{-2\frac{\ln(2/\delta)}{2\epsilon^2}\epsilon^2} \\
 &\leq 2e^{-\ln(2/\delta)} \\
 &\leq 2e^{\ln(\delta/2)} \\
 &\leq 2 \cdot \delta/2 \\
 &\leq \delta
 \end{aligned}$$

$$\text{Fix } s = \frac{\ln(2/\delta)}{2\epsilon^2}$$

## Lecture 2 Connectivity

$G$  is  $\epsilon$ -close to connected if you can modify at most  $\epsilon n$  entries in the adjacency list to make it connected.  
 If  $G(V, E)$  is connected, output True, else if  $\epsilon$ -far from connected, output False.

```

Connected(G, n, d, ε)
  Repeat 16/εd times:
    - Choose a random node u
    - Do a BFS from u, stopping after 8/εd nodes seen.
    - If CC of u has ≤ 8/εd nodes, return FALSE.
  return TRUE
  
```

BFS cost =  $\frac{8}{\epsilon d} \cdot d$   
 Total complexity =  $O(\frac{1}{\epsilon^2 d})$

## Lecture 2 Connected Components

Output CC such that:  $|CC(G) - C| \leq \epsilon n$ , w.p.  $> 1 - \delta$   
 Define  $n(u)$  = num nodes in the CC of  $u$ .  
 $cost(u) = 1/n(u)$ .  
 $\sum_{u \in CC_i} cost(u) = 1$ .

```

sum = 0
for j = 1 to s:
  - Sample u randomly
  - BFS from u, stop when see up to 2/ε nodes
  - If BFS found > 2/ε nodes:
    - sum := sum + ε/2
  - Else:
    - sum := sum + cost(u)
return n * (sum/s)
  
```

Let  $\bar{C} = \sum cost(u_j)$   
 Let  $Y_j = cost(u_j)$  of our sample  $j$   
 $|CC(G) - \bar{C}| \leq \epsilon n/2$   
 $E[Y_j] = \sum \frac{1}{n} cost(u_i) = \frac{1}{n} \bar{C}$   
 $E[\sum Y_j] = s \cdot E[Y_j] = \frac{s}{n} \bar{C}$   
 Since we output  $\frac{n}{s} \sum Y_j$ , we get  $E[\frac{n}{s} \sum Y_j] = \bar{C}$

$$\begin{aligned}
 P(|\bar{C} - \frac{n}{s} \sum Y_j| > \epsilon n/2) \\
 &= P(|E[\sum Y_j] - \sum Y_j| > \frac{s}{n} \epsilon n/2) \\
 &= P(|E[\sum Y_j] - \sum Y_j| > s\epsilon/2) \\
 &\leq 2e^{-2\epsilon^2 s^2 / 4s} \\
 &\leq 2e^{-\epsilon^2 s/2} \leq \delta
 \end{aligned}$$

Set  $s = \frac{2}{\epsilon^2} \ln(2/\delta)$ .  
 Complexity =  $2d/\epsilon \cdot O(\frac{1}{\epsilon^2} \ln(1/\delta)) = O(\frac{d}{\epsilon^3} \ln(1/\delta))$ .

### Lecture 3 MST weight

Output  $M$  such that:  $M = MST(G)(1 \pm \epsilon)$  w.p.  $> 1 - \delta$ .  
 Let  $G_j$  be the graph with edge weights  $j$  and below.  
 Let  $C_j$  be the connected components in  $G_j$ .  
 $MST(G)$  contains  $C_j - 1$  edges of weight  $> j$ .  
 Therefore:

$$MST(G) = (n - C_1) + \sum_{j=1}^{W-1} (j+1)(C_j - C_j + 1)$$

$$= n - W + \sum_{j=1}^{W-1} C_j$$

```
sum := n - W
for j = 1 to W - 1:
    -  $X_j = \text{ApproxCC}(G_j, d, \epsilon' \delta')$ 
    - sum := sum +  $X_j$ 
return sum
```

Sum of errors:  $(\epsilon' n)(W - 1)$   
 Set  $\epsilon' = \epsilon/W$ , then sum of errors  $\leq \epsilon n$ .  
 Set  $\delta' = \delta/W$ . Then  $P(\text{any fail}) \leq \sum_{j=1}^{W-1} \delta/W \leq \delta$ .  
 $MST(G) - \epsilon n \leq \text{sum} \leq MST(G) + \epsilon n$   
 Since  $MST(G) \geq n - 1 \geq n/2$ ,  $n \leq 2MST(G)$ ,  
 $MST(G)(1 - 2\epsilon) \leq \text{sum} \leq MST(G)(1 + 2\epsilon)$

### Lecture 3 Maximal Matching

Return the size of the Maximal Matching.

```
query(e):
    for all neighbours  $e'$  of  $e$ :
        if hash( $e'$ ) < hash( $e$ )
            if query( $e'$ ) = TRUE
                return FALSE
    return TRUE

sum := 0
for j = 1 to s:
    - Choose edge  $e$  uniformly at random.
    - if (query( $e$ ) = True)
        sum := sum + 1
return m.sum/s
```

$$E[\text{cost}] = 2 \sum_{k=1}^{\infty} \frac{d^k}{k!} = O(e^d).$$

$$\text{sum} = MM(G) \pm \epsilon m.$$

$$\text{Complexity} = O\left(\frac{e^d}{\epsilon^2}\right).$$

Can do better:  $O(\frac{d^4}{\epsilon^2})$ , even  $O(\frac{d^2}{\epsilon^2})$ , and reduce error to  $\pm \epsilon n$ .

### Lecture 3 Yao's Mini-Max

Every randomized algorithm on a worst-case input is always slower than the best deterministic algorithm on the worst distribution.

$$\forall A \in R : \max_{x \in X} (E[\text{cost}(A, x)]) \geq \min_{B \in D} (E[\text{cost}(B, x \text{ chosen from } \gamma)])$$

Recipe:

- Choose distribution  $\gamma$ .
- Show that the expected cost of every deterministic algorithm from  $\gamma$  is greater than some cost  $c$ .
- Conclude that every randomized algorithm has at least one input with expected cost at least as bad as  $c$ .

### Lecture 4 Misra Gries

$\text{count}(x)$ :  $N(x) - \epsilon m \leq \text{count}(x) \leq N(x) + \epsilon m$ .

Heavy Hitters: returns

- every item that appears  $\geq 2\epsilon m$  times.
- no item that appears  $< \epsilon m$  times.

```
Set  $P$  of  $\langle \text{item}, \text{count} \rangle$  pairs
For each  $u$  in stream  $S$ :
    1. if  $\langle u, c \rangle$  is in set  $P$ , increment  $c$ .
    2. else add  $\langle u, 1 \rangle$  to set  $P$ .
    3. if  $|P| > k$ , decrement count  $c$  for each item.
    4. Remove all items from  $P$  with count  $c = 0$ .
```

```
Count( $x$ ):
    1. if  $\langle x, c \rangle$  is in  $P$ , return  $c$ .
    2. else return 0.
```

Choose  $k = 1/\epsilon$ . Then  $N(x) - \epsilon m \leq \text{count}(x) \leq N(x)$ .

Space required:  $O(k \log m)$ .

Proof:

- Count of  $x$  is incremented  $N(x)$  times in total.
- Total increments is  $m$ .
- When  $\text{count}(x)$  is decremented, at least  $k$  items are also decremented.
- At most  $m$  decrements in total.
- So  $\text{count}(x)$  is decremented at most  $m/k$  times.

For Heavy Hitters: return  $x$  if  $\text{count}(x) \geq \epsilon m$ .

### Lecture 7 External Memory Model

Cache size =  $M$ . Block size =  $B$ . Number of lines (cache slots) =  $M/B$ .

Assumptions:

- One cache level.
- Only memory access has cost.
- Ideal cache and replacement.

Problem	Classical	EMM
Scan Linked List	$O(N)$	$O(N)$
Scan Array	$O(N)$	$O(N/B)$
Search Linked List	$O(N)$	$O(N)$
Search Red-black tree	$O(\log N)$	$O(\log N)$
Search Array	$O(\log N)$	$O(\log(N/B))$
Search B-tree	$O(\log_B N)$	$O(\log_B N)$
Sort B-tree	$O(N \log N)$	$O(N \log_B N)$
Read/Write B-tree	$O(\log_B N)$	$O(\log_B N)$

### Lecture 7 Caching

$(a, b)$  - tree with  $n$  keys has height  $\leq \log_a(\frac{n}{a}) + 1$ .

At most  $n/a$  leaves. Every node except the root has degree  $> a$ .

Node at height  $\log_a(\frac{n}{a})$  has  $\geq a^{\log_a(\frac{n}{a})} \geq \frac{n}{a}$  leaves.

Corollary: if  $a \geq B$ , then  $(a, b)$  - tree with  $n$  keys has height  $O(\log_B n)$ .

Amortized cost of split/share/merge is  $O(1/B)$ , thus  $O((1/B) \log_B N)$  per operation.

With parent points: insert may cost  $O(B \log_B N)$  if every level needs to split.