

1) <sup>a</sup> To prove:  $E[2^{C_n}] = n+1$ .

$$E[2^{C_n}] = \sum_{1 \leq k \leq n} 2^k \Pr(C_n = k)$$

$$\text{and } \Pr(C_n = k) = \frac{1}{2^{k-1}} \Pr(C_{n-1} = k-1) + \left(1 - \frac{1}{2^k}\right) \Pr(C_{n-1} = k) \dots \textcircled{1}$$

$$\begin{aligned} E[2^{C_n}] &= \sum_{1 \leq k \leq n} \left( 2 \Pr(C_{n-1} = k-1) + (2^k - 1) \Pr(C_{n-1} = k) \right) \\ &= \sum_{1 \leq k \leq n} \left( 2 \Pr(C_{n-1} = k-1) - \Pr(C_{n-1} = k) + 2^k \Pr(C_{n-1} = k) \right) \end{aligned}$$

using,

$$\sum_{1 \leq k \leq n} \Pr(C_n = k) = 1 \text{ and } \Pr(C_{n-1} = n) = 0$$

we get,

$$\begin{aligned} E[2^{C_n}] &= 2 - 1 + \sum_{1 \leq k \leq n-1} 2^k \Pr(C_{n-1} = k) \quad (\because \Pr(C_{n-1} = n) = 0) \\ &\quad + \Pr(C_{n-1} = 0) - 2^0 \Pr(C_{n-1} = 0) \\ &= E[2^{C_n}] = 1 + E[2^{C_{n-1}}] \end{aligned}$$

Base case:

$$E[2^{C_1}] = 2 \cdot \Pr(C_1 = 1) = \underline{2} \quad (\because \Pr(C_1 = 1) = \frac{1}{2^0} = 1)$$

$$\Rightarrow E[2^{C_n}] = n+1.$$

$$E[2^{2c_n}] = \sum_{1 \leq k \leq n} 2^{2k} \Pr(C_n = k) \quad \text{using part (1),}$$

$$= \sum_{1 \leq k \leq n} \left( 2^{k+1} \Pr(C_{n-1} = k-1) - 2^k \Pr(C_{n-1} = k) + 2^{2k} \Pr(C_{n-1} = k) \right)$$

$$= 2^2 E[2^{C_{n-1}}] - 1 E[2^{C_{n-1}}] + 2^0 \Pr(C_{n-1} = 0) - 2^0 \Pr(C_{n-1} = 0) + E[2^{2C_{n-1}}]$$

$$\text{E: } \Pr(C_{n-1} = n) = 0 \quad (\because \Pr(C_{n-1} = n) = 0)$$

$$\Rightarrow E[2^{2c_n}] = 3 E[2^{C_{n-1}}] + E[2^{2C_{n-1}}]$$

$$= 3n + E[2^{2C_{n-1}}]$$

Base case:

$$E[2^{2c_1}] = 2^2 \cdot \Pr(C_1 = 1) = \underline{4}.$$

$$\Rightarrow E[2^{2c_n}] = 3n + 3(n-1) + \dots + (3+1)$$

$$= 3 \left( \frac{n(n+1)}{2} \right) + 1$$

$$\text{Var}(2^{c_n}) = E[2^{2c_n}] - (E[2^{c_n}])^2$$

$$= 3 \left( \frac{n(n+1)}{2} \right) + 1 - (n+1)^2$$

$$= \frac{3}{2} (n^2 + n) + 1 - (n^2 + 1 + 2n)$$

$$= \frac{n^2 - n}{2} = \frac{n(n-1)}{2}$$

2) Algo

1) choose  $t$  hash functions  $h_1, \dots, h_t: \{1, 2, \dots, m\} \rightarrow \{1, 2, \dots, k\}$

each pairwise independent hash family

2) For  $1 \leq i \leq t$  &  $1 \leq j \leq k$  initialize counters  $C[i, j]$  to zero.  
The counters  $C[i, 1], \dots, C[i, k]$  correspond to  $i$ th hash function.

3) For each tuple  $(x_i, a(x_i))$  in stream, do:  
Set  $i_1 = h_1(x_i), \dots, i_t = h_t(x_i)$ .  
Increment each of  $C[1, i_1], C[2, i_2], \dots, C[t, i_t]$  by  $a(x_i)$ .

4) For each  $y \in \{1, \dots, m\}$ . output  $\{C[1, h_1(y)], C[2, h_2(y)], \dots, C[t, h_t(y)]\}$   
as estimated frequency defined.

3) divide the hash functions into two halves. first  $\lceil t/2 \rceil$  for positive  $a(x_i)$  and other half  $t - \lceil t/2 \rceil$  for negative  $a(x_i)$ .

4) For each tuple  $(x_i, a(x_i))$  in stream do:  
if  $a(x_i) > 0$   
Set  $i_1 = h_1(x_i), \dots, i_{\lceil t/2 \rceil} = h_{\lceil t/2 \rceil}(x_i)$   
Increment each of  $C[1, i_1], C[2, i_2], \dots, C[\lceil t/2 \rceil, i_{\lceil t/2 \rceil}]$   
by  $a(x_i)$ .

else  
Set  $i_{\lceil t/2 \rceil + 1} = h_{\lceil t/2 \rceil + 1}(x_i), \dots, i_t = h_t(x_i)$   
Increment each of  $C[\lceil t/2 \rceil + 1, i_{\lceil t/2 \rceil + 1}(x_i)], \dots, C[t, i_t(x_i)]$   
by  $-a(x_i)$ .

5) For each  $y \in \{1, \dots, m\}$  output  $\min \{C[1, h_1(y)], \dots, C[\lceil t/2 \rceil, h_{\lceil t/2 \rceil}(y)],$   
 $- \min \{C[\lceil t/2 \rceil + 1, h_{\lceil t/2 \rceil + 1}(y)], \dots, C[t, h_t(y)]\}$ .  
as estimated frequency defined.



Now, let  $F_P(y) = \sum_{y \in \text{stream}} a(y)$ , where all such  $a(y) > 0$

$F_N(y) = - \sum_{y \in \text{stream}} a(y)$ , where all such  $a(y) < 0$

Now, Fix  $y \in \{1, \dots, m\}$  where  $a(y) > 0$  &  $i \in \{1, \dots, \lceil t/2 \rceil\}$

$$\mathbb{E}[c(i, h_i(y))] = f(y) + \frac{F_P(y) - f(y)}{k} \leq f(y) + \frac{F_P(y)}{k}$$

(where  $f(y)$  is  $\sum_{i \in X} a(x_i)$ ,  $X$  is all such occurrence of  $x$ .)

Similarly for negative part,

$y \in \{1, \dots, m\}$  where  $a(y) < 0$  &  $j \in \{\lceil t/2 \rceil + 1, \dots, t\}$ .

$$\mathbb{E}[c(j, h_j(y))] = f(y) + \frac{F_N(y) - f(y)}{k} \leq f(y) + \frac{F_N(y)}{k}$$

(where  $f(y)$  is  $-\sum_{i \in X} a(x_i)$ ,  $X$  is occurrence of all such  $x$ .)

Let  $k = \lceil \frac{2}{\epsilon} \rceil$ . Using Markov's inequality,

$$\Pr(c[i, h_i(y)] - f(y) \geq \epsilon F_P(y)) \leq \frac{1}{\epsilon k} \leq \frac{1}{2}$$

$$\Pr\left(\min_{1 \leq k \leq \lceil t/2 \rceil} (c[k, h_k(y)] - f(y)) \geq \epsilon F_P(y)\right) \leq \frac{1}{2^{\lceil t/2 \rceil}} \dots \textcircled{1}$$

Similarly for negative part,

$$\Pr\left(\min_{\lceil t/2 \rceil + 1 \leq k \leq t} (c[k, h_k(y)] - f(y)) \geq \epsilon F_N(y)\right) \leq \frac{1}{2^{t - \lceil t/2 \rceil}} \approx \frac{1}{2^{\lceil t/2 \rceil}} \dots \textcircled{2}$$

~~Actual Value (x)~~

we need to find error,

$$\approx | \text{output}(x) - \text{actual}(x) |$$

$$\leq \left| \min_{\text{counter}}(\text{Pos}(x)) - \min_{\text{counter}}(\text{Neg}(x)) - (\text{Pos sum}(x) + \text{neg sum}(x)) \right|$$

$$\leq \left| \min_{\text{counter}}(\text{Pos}(x)) - \min_{\text{counter}}(\text{Neg}(x)) - (\text{Pos sum}(x) - (-\text{neg sum}(x))) \right|$$

(where, Pos counter is over  $C[]$  for value of Pos sum is  $f_p$ , neg sum is  $f_N$ )  
 Let  $f_k$  be sum of all  $a(x_i)$  in stream.

$$\Rightarrow \text{Prob}(|\text{output}(x) - \text{actual}(x)| \geq \epsilon(f_k))$$

$$\leq \text{Prob}(\min_{\text{counter}}(\text{Pos}(x)) - \text{Pos sum}(x) \geq \epsilon F_p(x))$$

$$+ \text{Prob}(\min_{\text{counter}}(\text{Neg}(x)) - (-\text{neg sum}(x)) \geq \epsilon F_N(x))$$

~~Prob~~

Using ① and ②,

$$\leq \frac{2}{2^{\lceil t/2 \rceil}} \leq \frac{2}{2^{t/2}} \leq \frac{1}{m^{1+c}} \quad \left( \text{let } t = \lceil 2(1+c) \log_2 m \rceil \right)$$

$$\text{for } c = \frac{1}{\log m} \log\left(\frac{1}{\delta}\right)$$

~~Let  $Z = \text{Max}(F_p, F_N)$~~

we get error to be at most  $\delta$ .  
 after applying union bound on  $m$  elements  
 as seen in lecture notes.

$$\text{Let } Z = \text{Max}(F_p, F_N) \cdot \text{Space comp} = O\left(\frac{\log(Z)}{\epsilon} (\log m + \log\left(\frac{1}{\delta}\right))\right)$$

### 3) Algorithm

1) Let  $E_A(i)$  be expected ~~no~~ of edges in cut when  $v_i$  is placed in  $A$ , and similarly  $E_B(i), E_C(i)$ .

2)  $E_A(i) = m_0 + d_B + d_C + (d_D \times 2/3)$  where  $m_0$  is no of edges currently in cut  
 $d_B =$  ~~edge~~ neighbours of  $v_i$  currently in  $B$ , similarly  $d_C$  and  $d_D$  is unplaced neighbours of  $v_i$ .

set  $A = \emptyset, B = \emptyset, C = \emptyset$

for  $i = 1$  to  $n$ , do:

find number of current neighbours of  $v_i$  in  $A, B, C$  respectively.  
 call them  $d_A, d_B, d_C$ .

if  $d_B + d_C > d_A + d_C$  and  $d_B + d_C > d_A + d_B$ , set  $A = A \cup \{v_i\}$

else if  $d_A + d_C > d_B + d_C$  and  $d_A + d_C > d_A + d_B$ , set  $B = B \cup \{v_i\}$

else set  $C = C \cup \{v_i\}$

output  $A, B, C$ .

Let  $\mu(i)$  be expected number of edges in cut after placing first  $i$  vertices.

~~$\mu(0) =$~~

Let  $H_e = \begin{cases} 1, & \text{edge } (u, v) \text{ s.t. } u \text{ and } v \text{ are in different set} \\ 0, & \text{else} \end{cases}$

$$\Rightarrow \mu(0) = \sum_{e \in E} H_e = \sum_{e \in E} \Pr(u, v \text{ are in different set})$$

$$= \frac{2m}{3}, \text{ where } m = |E|, \Pr(u, v \text{ in diff set}) = \frac{2}{3}.$$

$$\text{Also, } \mu(i-1) = \frac{1}{3} \left( \mu(i | v_i \in A) + \mu(i | v_i \in B) + \mu(i | v_i \in C) \right)$$

$$\& \mu(i) = \max \left( \mu(i | v_i \in A), \mu(i | v_i \in B), \mu(i | v_i \in C) \right)$$

$$\text{as } \max \geq \text{avg} \Rightarrow \mu(i) \geq \mu(i-1). \Rightarrow \text{for any } i, \mu(m) \geq \frac{2m}{3}.$$