CSCI B609: "Foundations of Data Science"

Lecture 19: L_0 -sampling, L_1 -sparse recovery, Count Sketch

Slides at http://grigory.us/data-science-class.html

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Data Streams

• Stream: m elements from universe $[n] = \{1, 2, ..., n\}$, e.g.

$$\langle x_1, x_2, ..., x_m \rangle = \langle 5, 8, 1, 1, 1, 4, 3, 5, ..., 10 \rangle$$

• f_i = frequency of i in the stream = # of occurrences of value i

$$f = \langle f_1, \dots, f_n \rangle$$

Frequency Moments

- Define $F_k = \sum_i f_i^k$ for $k \in \{0,1,2,...\}$
 - $-F_0 = \#$ number of distinct elements
 - $-F_1 = \#$ elements
 - $-F_2$ = "Gini index", "surprise index"

Proof of Lemma

- Let $j = \lceil \log \widetilde{F_0} \rceil$ and note that $2F_0 < 2^j < 12 F_0$
- For any i, $\Pr[h_j(i) = 0] = \frac{1}{2^j}$
- Probability there exists a unique i such that $h_i(i) = 0$,

$$\sum_{i} \Pr[h_{j}(i) = 0 \text{ and } \forall k \neq i, h_{j}(k) \neq 0]$$

$$= \sum_{i} \Pr[h_{j}(i) = 0] \Pr[\forall k \neq i, h_{j}(k) \neq 0 | h_{j}(i) = 0]$$

$$\geq \sum_{i} \Pr[h_{j}(i) = 0] \left(1 - \sum_{k \neq i} \Pr[h_{j}(k) = 0 | h_{j}(i) = 0]\right)$$

$$= \sum_{i} \Pr[h_{j}(i) = 0] \left(1 - \sum_{k \neq i} \Pr[h_{j}(k) = 0]\right) \geq \sum_{i} \frac{1}{2^{j}} \left(1 - \frac{F_{0}}{2^{j}}\right) \geq \frac{1}{24}$$

• Holds even if h_i are only 2-wise independent

Sparse Recovery

- Goal: Find g such that $||f g||_1$ is minimized among g's with at most k non-zero entries.
- Definition: $Err^k(f) = \min_{g:||g||_0 \le k} ||f g||_1$
- Exercise: $Err^k(f) = \sum_{i \notin S} |f_i|$ where S are indices of k largest f_i
- Using $O(\epsilon^{-1}k\log n)$ space we can find \tilde{g} such that $||\tilde{g}||_0 \le k$ and

$$|\tilde{g} - f||_1 \le (1 + \epsilon) Err^k(f)$$

Count-Min Revisited

- Use Count-Min with $d = O(\log n)$, $w = 4k/\epsilon$
- For $i \in [n]$, let $\tilde{f}_i = c_{j,h_i(i)}$ for some row $j \in [d]$
- Let $S = \{i_1, ..., i_k\}$ be the indices with max. frequencies. Let A_i be the event there doesn't exist $k \in S/i$ with $h_i(i) = h_i(k)$
- Then for $i \in [n]$:

$$\Pr\left[\left|f_{i}-\widetilde{f}_{i}\right| \geq \frac{\epsilon Err^{k}(f)}{k}\right] =$$

$$\Pr\left[\operatorname{not} A_{i}\right] \times \Pr\left[\left|f_{i}-\widetilde{f}_{i}\right| \geq \frac{\epsilon Err^{k}(f)}{k} \middle| \operatorname{not} A_{i}\right] +$$

$$\Pr\left[A_{i}\right] \times \Pr\left[\left|f_{i}-\widetilde{f}_{i}\right| \geq \frac{\epsilon Err^{k}(f)}{k} \middle| A_{i}\right]$$

$$\leq \Pr\left[\operatorname{not} A_{i}\right] + \Pr\left[\left|f_{i}-\widetilde{f}_{i}\right| \geq \frac{\epsilon Err^{k}(f)}{k} \middle| A_{i}\right] \leq \frac{k}{w} + \frac{1}{4} \leq \frac{1}{2}$$

• Because $d = O(\log n)$ w.h.p. all f_i 's approx . up to $\frac{\epsilon Err^k(f)}{k}$

Sparse Recovery Algorithm

- Use Count-Min with $d = O(\log n)$, $w = 4k/\epsilon$
- Let $f' = (\widetilde{f}_1, \widetilde{f}_2, ..., \widetilde{f}_n)$ be frequency estimates:

$$\left|f_i - \widetilde{f}_i\right| \le \frac{\epsilon Err^k(f)}{k}$$

- Let \tilde{g} be f' with all but the k-th largest entries replaced by 0.
- Lemma: $||\tilde{g} f||_1 \le (1 + 3\epsilon)Err^k(f)$

$$\left| \left| \tilde{g} - f \right| \right|_1 \le (1 + 3 \epsilon) Err^k(f)$$

- Let $S, T \subseteq [n]$ be indices corresponding to k largest values of f and f'.
- For a vector $x \in \mathbb{R}^n$ and $I \subseteq [n]$ denote as x_I the vector formed by zeroing out all entries of x except for those in I.

$$\begin{aligned} \left| |f - f_{T}'| \right|_{1} &\leq \left| |f - f_{T}| \right|_{1} + \left| |f_{T} - f_{T}'| \right|_{1} \\ &= \left| |f| \right|_{1} - \left| |f_{T}| \right|_{1} + \left| |f_{T} - f_{T}'| \right|_{1} \\ &= \left| |f| \right|_{1} - \left| |f_{T}'| \right|_{1} + \left(\left| |f_{T}'| \right|_{1} - \left| |f_{T}| \right|_{1} \right) + \left| |f_{T} - f_{T}'| \right|_{1} \\ &\leq \left| |f| \right|_{1} - \left| |f_{T}'| \right|_{1} + 2 \left| |f_{T} - f_{T}'| \right|_{1} \\ &\leq \left| |f| \right|_{1} - \left| |f_{S}'| \right|_{1} + 2 \left| |f_{T} - f_{T}'| \right|_{1} \\ &\leq \left| |f| \right|_{1} - \left| |f_{S}| \right|_{1} + \left(\left| |f_{S}| \right|_{1} - \left| |f_{S}'| \right|_{1} \right) + 2 \left| |f_{T} - f_{T}'| \right|_{1} \\ &\leq \left| |f - f_{S}| \right|_{1} + \left| |f_{S} - f_{S}'| \right|_{1} + 2 \left| |f_{T} - f_{T}'| \right|_{1} \\ &\leq Err^{k}(f) + k \epsilon \frac{Err^{k}(f)}{k} + 2k \epsilon \frac{Err^{k}(f)}{k} \\ &\leq (1 + 3 \epsilon)Err^{k}(f) \end{aligned}$$

Count Sketch [Charikar, Chen, Farach-Colton]

• In addition to $H_i:[n] \to [w]$ use random signs $r_i[n] \to \{-1,1\}$

$$c_{i,j} = \sum_{x:H_i(x)=j} r_i(x) f_x$$

Estimate:

$$\hat{f}_x = median(r_1(x)c_{1,H_1(x)}, ..., r_d(x)c_{d,H_d(x)})$$

- Parameters: $d = O\left(\log \frac{1}{\delta}\right)$, $w = \frac{3}{\epsilon^2}$ $\Pr[|\widetilde{f}_x f_x| + \epsilon ||f||_2] \ge 1 \delta$
- Lemma: $E[r_i(x)c_{i,H_i(x)}] = f_x$
- Lemma: $Var[r_i(x)c_{i,H_i(x)}] \leq \frac{F_2}{w}$
- By Chebyshev: $\Pr[|r_i(x)c_{i,H_i(x)} f_x| \ge \epsilon \sqrt{F_2}] \le 1/3$
- By Chernoff with $d = O\left(\log \frac{1}{\delta}\right)$ error prob. 1δ .

Count Sketch Analysis

• Fix i and x. Let $X_y = I[H(x) = H(y)]$: $r(x)C_{H(x)} = \sum_{v} r(x)r(y)f_yX_y$

- Lemma: $E[r_i(x)c_{i,H_i(x)}] = f_x$ $E[r(x)C_{H(x)}] = E[f_x + \sum_{y \neq x} r(x)r(y)f(y)X_y] = f_x$
- Lemma: $Var[r_i(x)c_{i,H_i(x)}] \le \frac{F_2}{w}$ $Var[r(x)C_{H(x)}] \le E[(\sum_y r(x)r(y)f_yX_y)^2]$ $= E[\sum_y f_y^2 X_y^2 + (\sum_{y \ne z} r(y)r(z)f_y f_z X_y X_z)]$ $= F_2/w$