

Sparse Recovery

Given linear sketch Ax

Find Z st. $\|Z\|_0 \leq K$, $\|x - Z\|_2$ is minimized

$$\text{minimum error } \text{err}_2^k := \left(\sum_{i \notin S} x_i^2 \right)^{1/2}$$

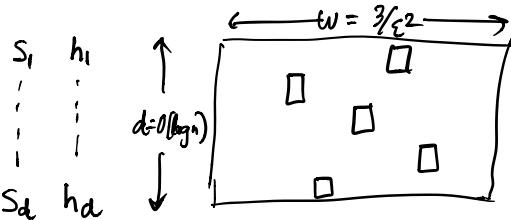
$S = \text{largest } K \text{ coordinates in } x$

Compressed Sensing

Count Sketch

$$\forall i \quad |\hat{x}_i - x_i| \leq \varepsilon \sqrt{F_2}$$

$$F_2 = \text{err}_2^0(x)$$



$$\text{Now: } w = \frac{3k}{\varepsilon^2}, \quad d = O(\log n)$$

$$\text{Lemma: W.h.p. } \forall i \quad |\hat{x}_i - x_i| \leq \frac{\varepsilon}{\sqrt{k}} \text{err}_2^k(x)$$

Proof. $\tilde{x}_{ij} : \text{estimate of } x_i \text{ from } j^{\text{th}} \text{ row } \{C_{ij}\}$

$$C_{jh_j(i)} \equiv C[j, h_j(i)]$$

$$\sigma_j(i) \equiv S_j(i)$$

$$\tilde{x}_{ij} = C_{jh_j(i)} \cdot \sigma_j(i)$$

$S : \text{largest } k \text{ coordinates of } x$

$A_i : \text{event that } \forall i' \in S \setminus i \text{ st. } h_j(i) = h_j(i')$

$$\begin{aligned} \Pr[|x_i - \tilde{x}_i| \geq \frac{\varepsilon}{\sqrt{k}} \text{err}_2^k] &= \Pr[A_i] \cdot \Pr[|x_i - \tilde{x}_i| \geq \frac{\varepsilon}{\sqrt{k}} \text{err}_2^k | A_i] \\ &\quad + \Pr[\bar{A}_i] \cdot \Pr[|x_i - \tilde{x}_i| \geq \frac{\varepsilon}{\sqrt{k}} \text{err}_2^k | \bar{A}_i] \\ &\leq \Pr[A_i] + \Pr[|x_i - \tilde{x}_i| \geq \frac{\varepsilon}{\sqrt{k}} \text{err}_2^k | \bar{A}_i] \end{aligned}$$

$$\Pr[A_i] \leq \frac{1}{w} \cdot K = \frac{\varepsilon^2}{3}$$

$$\mathbb{E}[|x_i - \tilde{x}_{ij}|^2 \mid \bar{A}_i] = \frac{(\text{err}_2^k(x))^2}{w}$$

Warning: I lied here!
The analysis of last time
does not work with
conditioning. Think about
how to fix
this issue.

$$\Pr[|x_i - \tilde{x}_{ij}| \geq \frac{\varepsilon}{\sqrt{K}} \text{err}_2^k(x) \mid \bar{A}_i] \leq \frac{1}{w} \cdot \frac{k}{\varepsilon^2} = \frac{1}{3}$$

$$\Pr[|x_i - \tilde{x}_{ij}| \geq \frac{\varepsilon}{\sqrt{K}} \text{err}_2^k(x)] \leq \frac{1}{3} + \frac{\varepsilon^2}{3}$$

Using median $\{\tilde{x}_{ij}\}$ we get w.h.p.

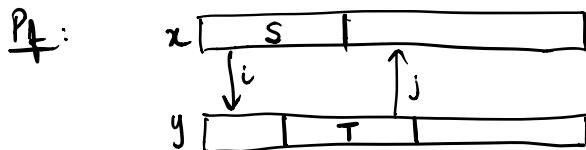
$$|\hat{x}_i - x_i| \leq \frac{\varepsilon}{\sqrt{K}} \text{err}_2^k(x)$$

Lemma: Given $\|x - y\|_\infty \leq \frac{\varepsilon}{\sqrt{K}} \text{err}_2^k(x)$

x : unknown vector
 y : count sketch estimates

T : largest K coordinates of y

$$Z = y_T \quad \|x - Z\|_2 \leq (1 + 5\varepsilon) \text{err}_2^k(x)$$



Break down error $\|x - Z\|_2^2$ into 3 terms

$$\|x - Z\|_2^2 = \|(x - Z)_T\|^2 + \|x_{S \setminus T}\|^2 + \|x_{[n] \setminus (S \cup T)}\|^2$$

$$\text{Let } E = \text{err}_2^k(x)$$

$$\|x - y\|_\infty \leq \frac{\varepsilon}{\sqrt{K}} \text{err}_2^k(x) = \frac{\varepsilon \cdot E}{\sqrt{K}}$$

$$\|(x - Z)_T\|^2 \leq K \cdot \frac{\varepsilon^2}{K} E^2 = \varepsilon^2 E^2$$

$$\forall i \in S \setminus T \quad |x_i| \leq |y_i| + \frac{\varepsilon}{\sqrt{K}} E$$

$$\forall j \in T \setminus S \quad -|x_j| \leq -|y_j| + \frac{\varepsilon}{\sqrt{K}} E \leq |x_j| \geq |y_j| - \frac{\varepsilon}{\sqrt{K}} E$$

$$|x_i| - |x_j| \leq |y_i| - |y_j| + \frac{2\varepsilon}{\sqrt{K}} E \leq \frac{2\varepsilon}{\sqrt{K}} E$$

$$|y_j| \geq |y_i|$$

Let $a = \max_{i \in S \setminus T} |x_i|$, $b = \min_{j \in T \setminus S} |x_j|$

$$a - b \leq \frac{2\epsilon}{\sqrt{K}} E$$

$$\begin{aligned} \|x_{S \setminus T}\|_2^2 &\leq a^2 |S \setminus T| \\ &\leq \left(b + \frac{2\epsilon}{\sqrt{K}} E\right)^2 |S \setminus T| \\ &\downarrow \\ &\leq \left(\frac{\|x_{T \setminus S}\|_2}{\sqrt{|S \setminus T|}} + \frac{2\epsilon}{\sqrt{K}} E\right)^2 |S \setminus T| \\ &\leq \left(\|x_{T \setminus S}\|_2 + 2\epsilon E\right)^2 \quad |S \setminus T| \leq K \\ &= \|x_{T \setminus S}\|_2^2 + 4\epsilon E \underbrace{\|x_{T \setminus S}\|_2}_{\leq E} + 4\epsilon^2 E^2 \\ &\leq \|x_{T \setminus S}\|_2^2 + 8\epsilon E^2 \quad (\epsilon^2 \leq \epsilon) \end{aligned}$$

$$\begin{aligned} \|x - z\|_2^2 &= \|(\bar{x} - z)_T\|_2^2 + \|x_{S \setminus T}\|_2^2 + \|x_{[n] \setminus (S \cup T)}\|_2^2 \\ &\leq \epsilon^2 E^2 + 8\epsilon E^2 + \underbrace{\|x_{T \setminus S}\|_2^2 + \|x_{[n] \setminus (S \cup T)}\|_2^2}_{\leq E^2} \end{aligned}$$

$$\|x_{T \setminus S}\|_2^2 + \|x_{[n] \setminus (S \cup T)}\|_2^2 = E^2$$

| | | | | |
|---|--|----------------|---|----------------|
| x | <table border="1"><tr> <td>S</td> <td>T</td> <td>S \ (S \cup T)</td> </tr></table> | S | T | S \ (S \cup T) |
| S | T | S \ (S \cup T) | | |
| y | <table border="1"><tr> <td></td> <td>T</td> <td></td> </tr></table> | | T | |
| | T | | | |

$$\|x - z\|_2^2 \leq \epsilon^2 E^2 + 8\epsilon E^2 + E^2 \leq (1 + 10\epsilon) E^2$$

$$\|x - z\|_2 \leq (1 + 5\epsilon) E$$

Next time: lo-sampling

Priority Sampling

Given stream of items with weights w_1, \dots, w_n

Want to store representative sample S of the items
with weights \hat{w}_i

Given query $I \subset [n]$, we would like to estimate

$$w_I = \sum_{i \in I} w_i \quad \text{by} \quad \sum_{i \in I \cap S} \hat{w}_i$$

1. For data w_1, \dots, w_n sample $u_i \in [0, 1]$ uniformly at random independently

$$2. \text{Priority } q_i = \frac{w_i}{u_i}$$

3. Keep set S_K of largest K priorities seen so far
as well as value τ' of $(K+1)^{\text{th}}$ largest priority

$$4. \hat{w}_i = \begin{cases} \max\{\tau', w_i\} & \text{if } i \in S_K \\ 0 & \text{otherwise} \end{cases}$$

Given I , output $\hat{w}_I = \sum_{j \in I \cap S_K} \hat{w}_j$

Lemma: $E[\hat{w}_i] = w_i$

Proof: Let $A(\tau')$ be event that $(K+1)^{\text{th}}$ highest priority is τ'

Given $A(\tau')$, for all $i \in S_K$, priority $q_i = \frac{w_i}{u_i} > \tau'$

$$\text{weight } \hat{w}_i = \max(\tau', w_i)$$

$$\text{if } i \notin S_K, \quad q_i < \tau', \quad \hat{w}_i = 0$$

$$P(i \in S | A(\tau'))$$

$$\textcircled{1} \quad w_i > \tau': \quad P(i \in S_K | A(\tau')) = 1, \quad \hat{w}_i = w_i \quad E[\hat{w}_i | A(\tau')] = w_i$$

$$\textcircled{2} \quad w_i \leq \tau': \quad P(i \in S_K | A(\tau')) = P(u_i \leq \frac{w_i}{\tau'}) = \frac{w_i}{\tau'}$$

$$\hat{w}_i = \tau'$$

$$E[\hat{w}_i | A(\tau')] = w_i$$

$$E[\hat{w}_i] = w_i$$

$$\text{Cov}[\hat{w}_i \hat{w}_j] = 0$$

Lemma $E[\hat{w}_i \hat{w}_j] = w_i w_j$

Lemma $E\left[\prod_{i \in I} \hat{w}_i\right] = \prod_{i=1}^K w_i \quad |I| \leq K$
 $= 0 \quad |I| > K$

Pf: $|I| \leq K$

Let τ'' be the $(K - |I| + 1)$ th highest priority among the elements $j \neq i$

$A(\tau'')$: corresponding event

$$E\left[\prod_{i \in I} \hat{w}_i \mid A(\tau'')\right] = \prod_{i \in I} w_i$$

Base case: $|I| = 1$

Case 1: $\exists h \in I \quad w_h > \tau''$ Clearly $h \in S_K$, $\hat{w}_h = w_h$

$$E\left[\prod_{i \in I} \hat{w}_i \mid A(\tau'')\right] = w_h \cdot E\left[\prod_{i \in I \setminus \{h\}} \hat{w}_i \mid A(\tau'')\right]$$

Apply induction to $I \setminus \{h\}$

Case 2: For all $h \in I, w_h < \tau''$

Let q be min priority among items in I

If $q < \tau''$, then $\hat{w}_j = 0$ for some $j \in I$
 \Rightarrow product will be 0

Consider case when $q > \tau''$

$\prod_{i \in I} \left(\frac{w_i}{\tau''}\right) \leftarrow$ Pr all elements $i \in I$ have priority $> \tau''$

$$\hat{w}_i = \tau''$$

$$E\left[\prod_{i \in I} \hat{w}_i\right] = \prod_{i \in I} (\tau'') \cdot \prod_{i \in I} \left(\frac{w_i}{\tau''}\right) = \prod_{i \in I} w_i$$