

Round-query tradeoff for Single Element Recovery on affine sets

Benjamin Rossman
Duke University

September 11, 2021

Abstract

Let \mathbb{F} be a field and A an affine subset of \mathbb{F}^N that does not contain the origin. How many linear queries $\mathbb{F}^N \rightarrow \mathbb{F}$ are required to find a nonzero coordinate of a hidden vector in A ? We show that the deterministic r -round query complexity of this problem is between $r(m(A)^{1/r} - 1)$ and $rm(A)^{1/r}$ where

$$m(A) := \min_{y \in \mathbb{F}^N : \forall x \in A, \langle x, y \rangle = 1} |\text{Supp}(y)|.$$

(In particular, $m(A) = N$ when $A = \{x \in \mathbb{F}^N : \sum_i x_i = 1\}$.) The proof is a direct generalization of the special case for \mathbb{F}_2 given in [Ros17].

1 Introduction

Let Γ be a set with a zero element 0 (for example: the set $\{0, 1\}$, any field \mathbb{F} , or the nonnegative real numbers $\mathbb{R}_{\geq 0}$). Let N be a positive integer, and let $[N] := \{1, \dots, N\}$.

For a vector $x = (x_1, \dots, x_N) \in \Gamma^N$, its *support* is the set $\{i \in [N] : x_i \neq 0\}$ of nonzero coordinates. We denote by $\vec{0}$ the all-zero vector in Γ^N with empty support.

For a set $\mathcal{D} \subseteq \Gamma^N$ and a set \mathcal{Q} of functions whose domain includes \mathcal{D} , we consider the following *Single Element Recovery* problem: For an arbitrary hidden vector $x \in \mathcal{D} \setminus \{\vec{0}\}$, find an element of $\text{Supp}(x)$ (i.e., a nonzero coordinate of x) by making queries $q_1, \dots, q_k \in \mathcal{Q}$ and learning answers $q_1(x), \dots, q_k(x)$. We consider the query complexity of this problem (i.e., minimum required number of queries) in different settings: randomized vs. deterministic, as well as non-adaptive vs. r -round adaptive.

For various settings of \mathcal{D} and \mathcal{Q} that have been studied, the following upper bounds are known to hold:

Randomized non-adaptive $O(\log^2 N)$ query algorithm. By the Valiant-Vazirani Isolation Lemma [VV85], there is a joint distribution of $t = O(\log N \cdot \log \frac{1}{\varepsilon})$ random sets $I_1, \dots, I_t \subseteq [N]$ such that, for every nonempty $S \subseteq [N]$,

$$\mathbb{P}_{I_1, \dots, I_t} [\exists j \in \{1, \dots, t\}, |S \cap I_j| = 1] \geq 1 - \varepsilon.$$

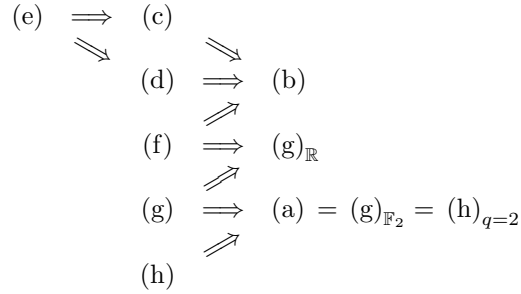
For each $j \in \{1, \dots, t\}$, the algorithm makes $O(\log N)$ queries that indicate whether $\text{Supp}(x) \cap I_j$ is a singleton and, if so, give the unique coordinate in this set.

Deterministic r -round $rN^{1/r}$ query algorithm. The algorithm maintains a nonempty set of coordinates, initially $[N]$ itself, that is known to contain a nonzero coordinate of x . In each round, the algorithm reduces the set to one of $\lceil N^{1/r} \rceil$ subsets in a balanced partition. (When $r = \lceil \log_2 N \rceil$, this is the usual binary search using r queries.)

The following table lists settings of \mathcal{D} and \mathcal{Q} where both upper bounds hold. The righthand columns gives references for matching lower bounds. Question marks indicate the lower bound is unknown (or the author is unaware of a reference).

\mathcal{D}	\mathcal{Q}	rand. non-adaptive $\Omega(\log^2 N)$ l.b.	deterministic r -round $\Omega(r(N^{1/r} - 1))$ l.b.
$\{x \in \{0, 1\}^N : x \text{ is odd}\}$	XOR ^(a)	?	[Ros17, Ros18]
$\{0, 1\}^N$	OR ^(b)	[JST11]	[ACK20]
	monotone ^(c)	[KRW17]	?
$\left\{x \in \{0, 1\}^{\mathbb{F}_2^{\log N}} : x \text{ is the char. vector of an affine subset of } \mathbb{F}_2^{\log N}\right\}$	OR ^(d)	[KRW17]	?
	monotone ^(e)	[Ros21]	?
$\mathbb{R}_{\geq 0}^N$	linear ^(f)	[JST11]	[ACK20]
$\{x \in \mathbb{F}^N : \sum_i x_i \neq 0\}$, any field \mathbb{F}	linear ^(g)	?	this note
$\left\{x \in (\mathbb{Z}/q\mathbb{Z})^N : \forall i, x_i \in \{0, 1\} \text{ and } \sum_i x_i \neq 0\right\}$, any $q \geq 2$	linear ^(h)	?	[CS21]

The diagram below shows implications between lower bounds from different rows of this table:



Most of these implications follow trivially from containments among the different classes \mathcal{D} and \mathcal{Q} : we get a stronger lower bound whenever we are able to shrink \mathcal{D} or augment \mathcal{Q} . Implication (f) \implies (b) follows from the observation that an OR query $\bigvee_{i \in I} x_i$ ($I \subseteq [N]$) on domain $\{0, 1\}^N$ carries less information than the corresponding linear query $\sum_{i \in I} x_i$. Note that the lower bounds in rows (e), (f) and (g) imply all other lower bounds in the above table.

Some comments on specific rows in the table:

Row (a). Any deterministic r -round q -query algorithm in this setting converts to a depth $r + 1$, size 2^{r+q} AC⁰ formula solving PARITY_N. Lower bounds $\Omega(N^{1/r})$ and $\Omega(r(N^{1/r} - 1))$ in row (a) follow from the AC⁰ formula lower bounds of [Ros18]. (A weaker $\Omega(N^{1/r})$ bound follows from the much earlier AC⁰ circuit lower bound of Håstad [Hås86].) [Ros17] subsequently strengthened the lower bound to $r(N^{1/r} - 1)$ (improving the previous big- Ω constant to the optimal value 1) by a linear-algebraic argument that avoids the random restriction based method of circuit complexity. (A different proof of the same $r(N^{1/r} - 1)$ lower bound appears in [CS21].) The lower bound for row (g) presented in this note is a direct generalization of the argument in [Ros17] from \mathbb{F}_2 to arbitrary fields.

Rows (d) and (e). Here we assume that $N = 2^n$ for some integer n , and we identify $\{0, 1\}^N$ with subsets of \mathbb{F}_2^n . Elements of \mathcal{D} thus correspond to affine subsets of \mathbb{F}_2^n , and Single Element Recovery may be viewed as the search problem of finding a vector in a hidden affine set. (The lower bound in [Ros21] is presented for the slightly different, but equivalent, search problem of finding a nonzero vector in a hidden non-trivial linear subspace of \mathbb{F}_2^n .)

Row (b). [ACK20] show that $N^{1/r} - 1$ queries are required in at least one round. A straightforward extension of their argument yields a stronger $r(N^{1/r} - 1)$ lower bound on the total number of queries.

Row (f). [ACK20] show that $\frac{1}{2}(N^{1/r} - 1)$ linear queries $\mathbb{R}^n \rightarrow \mathbb{R}$ are required in at least one round. It is again a straightforward extension to obtain a stronger $\frac{1}{2}r(N^{1/r} - 1)$ lower bound on the total number of queries. Here the factor $\frac{1}{2}$ arises from splitting each linear query into two nonnegative linear queries; it is unclear if this factor is necessary for domain $\mathbb{R}_{\geq 0}^N$. (If so, the implication (f) \Rightarrow (g) $_{\mathbb{R}}$ holds up to a factor 2.) The $r(N^{1/r} - 1)$ lower bound for row (g), proved in this note, shows that this factor is unnecessary for domain \mathbb{R}^N .

Rows (g) and (h). When q is prime, we may compare the r -round lower bounds of rows (g) and (h) for the field $\mathbb{F}_q = \mathbb{Z}/q\mathbb{Z}$. For $q = 2$, both (g) and (h) specialize to the $rN^{1/r} - 1$ bound of row (a). However, these bounds are incomparable for primes $q \geq 3$. Quantitatively, the r -round lower bound in row (g) is $rN^{1/r} - 1$ for all fields \mathbb{F} , while row (h) is $\Omega(\frac{r(N^{1/r}-1)}{q^{1+1/r}\log^2 q})$; for this reason, (h) $\not\Rightarrow$ (g). On the other hand, the domain of (h) is smaller than the domain of (g), as it includes the additional constraint that $x \in \{0, 1\}^N$; for this reason, (g) $\not\Rightarrow$ (h).

Definition 1. If A is an affine subset of \mathbb{F}^N that does not contain the origin, we define $m(A) \geq 1$ by

$$m(A) := \min_{y \in \mathbb{F}^N : \forall x \in A, \langle x, y \rangle = 1} |y|.$$

For example, if A is the codimension-1 affine set $\{x \in \mathbb{F}^N : \sum_i x_i = 1\}$, then we have $m(A) = N$.

Theorem 2. *Let A be an affine subspace of \mathbb{F}^N that does not contain the origin. Then the number of linear queries $\mathbb{F}^N \rightarrow \mathbb{F}$ required to solve Single Element Recovery on domain A is at least $r(m(A)^{1/r} - 1)$ and at most $rm(A)^{1/r}$.*

2 Round-query tradeoff over any field

Fix an arbitrary field \mathbb{F} and positive integer N . We consider the vector space \mathbb{F}^N . For $x \in \mathbb{F}^N$, let $|x|$ denote the number of nonzero coordinates of x (i.e., the size of support of x , not the sum of coordinates $\sum_i x_i$).

In what follows, U, V, S, T represent linear subspaces of \mathbb{F}^N , while A, B represent affine subsets of \mathbb{F}^N that do not contain the origin. Notation $U <_k V$ expresses that U is a codimension- k subspace of V .

Fix an inner product $\langle \cdot, \cdot \rangle$ on \mathbb{F}^N . Let V^\perp denote the orthogonal complement of a linear subspace V , that is, $V^\perp := \{x \in \mathbb{F}^N : \langle x, v \rangle = 0 \text{ for all } v \in V\}$. Note that $(V^\perp)^\perp = V$ and $U <_k V \Rightarrow V^\perp <_k U^\perp$.

Lemma 3. *Let A be an affine subset of \mathbb{F}^N that does not contain the origin. Let $U <_1 V$ be linear subspaces*

$$U := \{x_1 - x_2 : x_1, x_2 \in A\} \quad \text{and} \quad V := \{\lambda x : x \in A, \lambda \in \mathbb{F}\}.$$

Then $m(A) = \min_{z \in U^\perp \setminus V^\perp} |z|$.

Lemma 4. *For all $U <_k V \leq \mathbb{F}^N$, there exists a linear projection $\pi : V \rightarrow U$ such that $|\pi(v)| \leq (k+1)|v|$ for all $v \in V$.*

Proof. Let w_1, \dots, w_k be a basis of V over U , greedily chosen so that $|w_i|$ is maximal among elements $w_i \in V \setminus \text{LinearSpan}(U \cup \{w_1, \dots, w_{i-1}\})$ for each $i \in \{1, \dots, k\}$. Consider any $v \in V$. Then $v = u + a_1 w_1 + \dots + a_k w_k$ for some $u \in U$ and $a_1, \dots, a_k \in \mathbb{F}$. Let $\pi : V \rightarrow U$ be the projection $v \mapsto u$. Then

$$|\pi(v)| = |u| = |v - (a_1 w_1 + \dots + a_k w_k)| \leq |v| + |a_1 w_1| + \dots + |a_k w_k|.$$

For each i , we either have $a_i = 0$, in which case $|a_i w_i| = 0$; or $a_i \neq 0$, in which case $v \notin \text{LinearSpan}(U \cup \{w_1, \dots, w_{i-1}\})$ and therefore $|v| \leq |w_i| = |a_i w_i|$. We conclude that $|\pi(v)| \leq (k+1)|v|$ as required. \square

Definition 5. For real $r, m \geq 0$, let

$$\beta(r, m) := \lim_{\rho \searrow r} \rho(m^{1/\rho} - 1) = \begin{cases} 0 & \text{if } r = 0 \text{ and } m \leq 1, \\ \infty & \text{if } r = 0 \text{ and } m > 1, \\ r(m^{1/r} - 1) & \text{if } r > 0. \end{cases}$$

As an aside: note that $\lim_{r \rightarrow \infty} \beta(r, m) = \ln(m)$.

Lemma 6. For all real $r, m, k \geq 0$,

$$\beta(r, m/\ell) + \ell - 1 \geq \beta(r + 1, m).$$

When $r > 0$, this holds with equality if and only if $\ell = m^{1/(r+1)}$.

Proof. Elementary calculus. □

Theorem 7 (Lower Bound). Let A be an affine subspace of \mathbb{F}^N that does not contain the origin. Then r -round Single Element Recovery on domain A requires at least $\beta(r, m(A))$ linear queries.

Proof. We prove the theorem by induction on r . In the base case $r = 0$, the theorem states that Single Element Recovery on domain A is solved in zero rounds (with zero queries) if, and only if, $m(A) = 1$. This follows from the definition of $m(A)$.

For the induction step, let $r \geq 1$ and assume the theorem holds for fewer than r rounds. Suppose an optimal deterministic r -round algorithm makes k queries in the first round. These queries correspond to a linear function

$$\varphi : \mathbb{F}^N \rightarrow \mathbb{F}^k.$$

For each possible sequence of answers $b \in \mathbb{F}^k$, we are left with a deterministic $r-1$ -round algorithm that solves Single Element Recovery over $\{x \in A : \varphi(x) = b\}$, which is a codimension- k affine subset of A . We show how to adversarially choose an answer sequence $b \in \mathbb{F}^k$ such that the remaining $r-1$ -round algorithms requires at least $\beta(r-1, m(A)/(k+1))$ queries. The total number of queries is then at least $k + \beta(r-1, m(A)/(k+1))$, which is $\geq \beta(r, m(A))$ by Lemma 6.

We proceed to explain how to find a suitable answer sequence $b \in \mathbb{F}^k$ for linear queries $\varphi : \mathbb{F}^N \rightarrow \mathbb{F}^k$. As in Lemma 3, let $U <_1 V$ be linear subspaces

$$U := \{x_1 - x_2 : x_1, x_2 \in A\} \quad \text{and} \quad V := \{\lambda x : x \in A, \lambda \in \mathbb{F}\}.$$

Let

$$S := \text{Ker}(\varphi) \cap U.$$

Without loss of generality, we may assume that $S <_k U$, since otherwise a proper subset of the k queries would contain the same information about a hidden vector in A .

By duality, we have $V^\perp <_1 U^\perp <_k S^\perp$. By Lemma 4, there exists a linear projection

$$\pi : S^\perp \rightarrow U^\perp$$

such that $|\pi(z)| \leq (k+1)|z|$ for all $z \in S^\perp$. Let

$$T := (\pi^{-1}(V^\perp))^\perp.$$

By duality and the rank-nullity theorem, we have

$$T^\perp = \pi^{-1}(V^\perp), \quad V^\perp <_k T^\perp <_1 S^\perp, \quad S <_1 T <_k V, \quad S = T \cap U.$$

Note that $T \not\subseteq U$, since otherwise $S = T$ (whereas we have shown $S <_1 T$). Fix any $t \in T \setminus U$. Since $V = \{\lambda x : x \in A, \lambda \in \mathbb{F}\}$, there exist $x_0 \in A$ and $c \in \mathbb{F}$ such that $cx_0 = t$; moreover, $c \neq 0$ (since $t \neq \vec{0}$ as $t \notin U$). Now let

$$B := \{s + c^{-1}t : s \in S\}.$$

We claim that

$$B = \{x \in A : \varphi(x) = \varphi(t)\}.$$

We first prove the containment \subseteq . Consider any element $s + c^{-1}t \in B$ where $s \in S$. Since $c^{-1}t = x_0 \in A$ and $S \subseteq U = \{x_1 - x_2 : x_1, x_2 \in A\}$, we have $s + c^{-1}t = x_0 + x_1 - x_2$ where $x_0, x_1, x_2 \in A$. Therefore, $s + c^{-1}t \in A$. Having shown $B \subseteq \{x \in A : \varphi(x) = \varphi(t)\}$, it then follows that these sets are equal, since both are affine sets of dimension $\dim(A) - k$.

Note that

$$S = \{w_1 - w_2 : w_1, w_2 \in B\} \quad \text{and} \quad T = \{\lambda w : w \in B, \lambda \in \mathbb{F}\}.$$

We have

$$\begin{aligned} m(B) &= \min_{z \in S^\perp \setminus T^\perp} |z| && \text{(Lemma 3)} \\ &= \min_{z \in \pi^{-1}(U^\perp) \setminus \pi^{-1}(V^\perp)} |z| && \text{(definition of } \pi \text{ and } T) \\ &\geq \min_{z \in \pi^{-1}(U^\perp) \setminus \pi^{-1}(V^\perp)} \frac{|\pi(z)|}{k+1} && \text{(property of } \pi \text{ given by Lemma 4)} \\ &= \min_{y \in U^\perp \setminus V^\perp} \frac{|y|}{k+1} \\ &= \frac{m(A)}{k+1} && \text{(Lemma 3).} \end{aligned}$$

Let $\varphi(t) \in \mathbb{F}^k$ be the (adversarially chosen) sequence of answers to first-round queries $\varphi : \mathbb{F}^N \rightarrow \mathbb{F}^k$. We are left with an $r - 1$ round algorithm that solves Single Element Recovery on domain B . By the induction hypothesis and Lemma 6, the total number of queries over all r rounds is at least

$$k + \beta(r - 1, m(B)) \geq k + \beta(r - 1, \frac{m(A)}{k+1}) \geq \beta(r, m(A)),$$

which completes the proof. \square

Theorem 8 (Upper Bound). *Let A be an affine subspace of \mathbb{F}^N that does not contain the origin. Then for all $r \geq 1$, Single Element Recovery on domain A is solvable in r rounds using $rm(A)^{1/r}$ linear queries.*

Proof. Fix $y \in \mathbb{F}^N$ such that $|y| = m(A)$ and $\langle x, y \rangle = 1$ for all $x \in A$. Let $I = \{i \in [N] : y_i \neq 0\}$ (so, $|y| = |I| = m(A)$). Let $I = I_1 \sqcup \dots \sqcup I_k$ be partition of I into $k \leq \lceil m(A)^{1/r} \rceil$ blocks of size at most $m(A)^{(r-1)/r}$. For $j \in \{1, \dots, k\}$, let $y^{(j)}$ be the sub-vector of y with support I_j .

In the first round, the algorithm makes $k - 1$ queries that learn the values of $\langle x, y^{(1)} \rangle, \dots, \langle x, y^{(k-1)} \rangle$. From the answers, we learn some $j \in \{1, \dots, k\}$ and $\lambda \in \mathbb{F}^\times$ such that $\langle x, y^{(j)} \rangle = \lambda$. The algorithm then proceeds

to solve Single Element Recovery on the corresponding affine subset B of A . We have $m(B) \leq m(A)^{(r-1)/r}$, as witnessed by the fact that $\langle w, \lambda^{-1}y^{(j)} \rangle = 1$ for all $w \in B$.

If $r = 1$, we are done, since $|y^{(j)}| = m(A)^0 = 1$ and we have used $k - 1 = m(A) - 1$ queries. If $r \geq 2$, then by induction the total number of queries is at most

$$\begin{aligned} k - 1 + (r - 1)m(B)^{1/(r-1)} &\leq \lceil m(A)^{1/r} \rceil - 1 + (r - 1)(m(A)^{(r-1)/r})^{1/(r-1)} \\ &\leq m(A)^{1/r} + (r - 1)m(A)^{1/r} \\ &= rm(A)^{1/r}. \end{aligned} \quad \square$$

References

- [ACK20] Sepehr Assadi, Deeparnab Chakrabarty, and Sanjeev Khanna. Graph connectivity and single element recovery via linear and or queries. *arXiv preprint arXiv:2007.06098*, 2020.
- [CS21] Amit Chakrabarti and Manuel Stoeckl. The element extraction problem and the cost of determinism and limited adaptivity in linear queries. *arXiv preprint arXiv:2107.05810*, 2021.
- [Hås86] Johan Håstad. Almost optimal lower bounds for small depth circuits. In *Proc. 18th ACM Symposium on Theory of Computing*, pages 6–20, 1986.
- [JST11] Hossein Jowhari, Mert Sağlam, and Gábor Tardos. Tight bounds for lp samplers, finding duplicates in streams, and related problems. In *Proceedings of the thirtieth ACM SIGMOD-SIGACT-SIGART symposium on Principles of database systems*, pages 49–58, 2011.
- [KRW17] Akinori Kawachi, Benjamin Rossman, and Osamu Watanabe. The query complexity of witness finding. *Theory of Computing Systems*, 61(2):305–321, 2017.
- [Ros17] Benjamin Rossman. Subspace-invariant AC^0 formulas. In *Proc. 44th International Colloquium on Automata, Languages and Programming (ICALP)*, volume 80 of *LIPICs*, pages 93:1–11, 2017.
- [Ros18] Benjamin Rossman. The average sensitivity of bounded-depth formulas. *Computational Complexity*, 27(2):209–223, 2018.
- [Ros21] Benjamin Rossman. Thresholds in the lattice of subspaces of \mathbb{F}_q^n . In *Latin American Symposium on Theoretical Informatics*, pages 504–515. Springer, 2021.
- [VV85] Leslie G Valiant and Vijay V Vazirani. Np is as easy as detecting unique solutions. In *Proceedings of the 17th Annual ACM Symposium on Theory of Computing*, pages 458–463, 1985.