

Sparse Approximation, List Decoding, and Uncertainty Principles

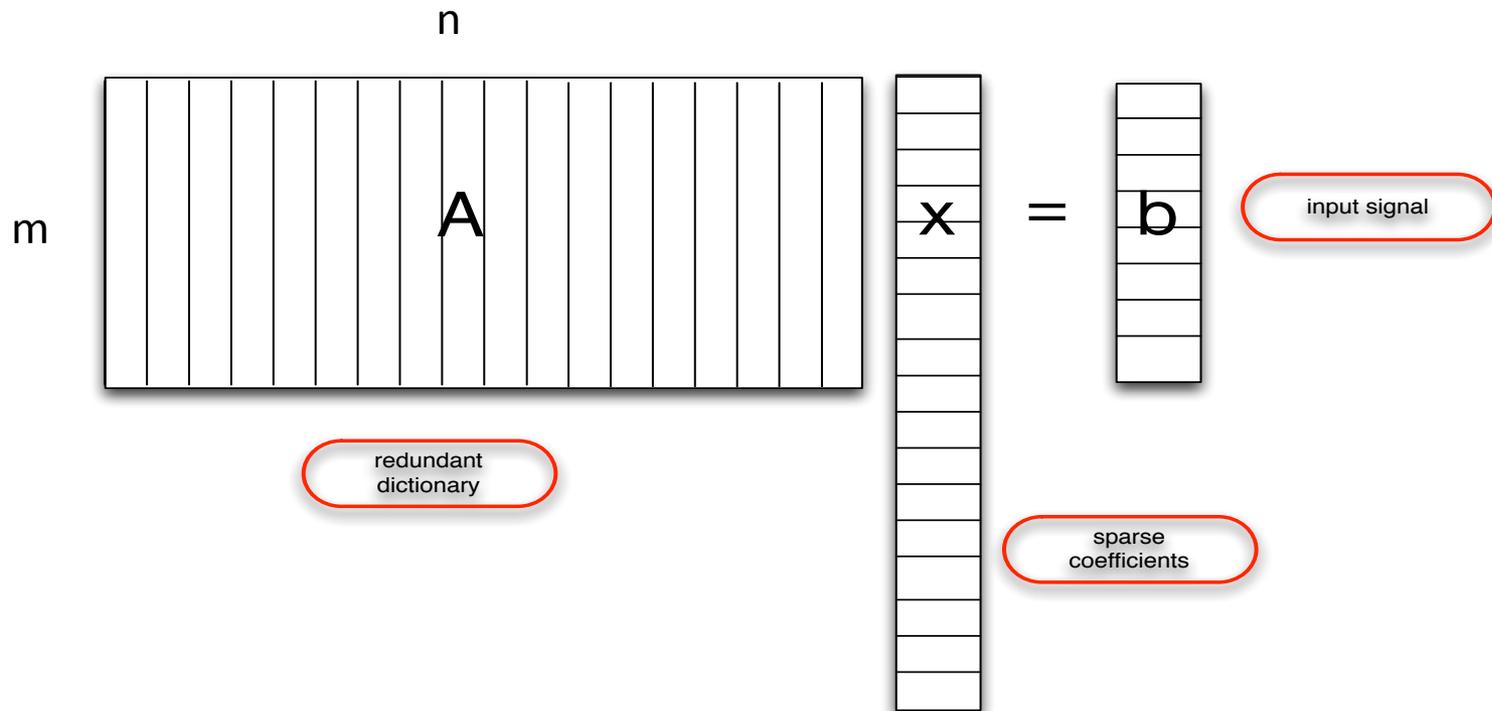
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Sparse Approximation: Definitions



EXACT: Given A, b , find sparsest x s.t. $Ax = b$.

$$\hat{x} = \operatorname{argmin} \|x\|_0 \text{ s.t. } Ax = b$$

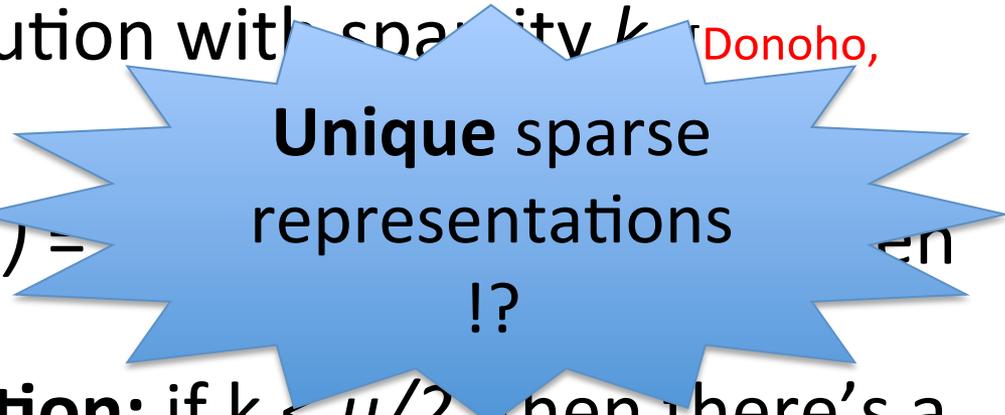
SPARSE: Given A, b, k , find best k -term approximation for b .

$$\hat{x} = \operatorname{argmin} \|Ax - b\|_2 \text{ s.t. } \|x\|_0 \leq k$$

Sparse Approximation: Unique repn.

“barriers”

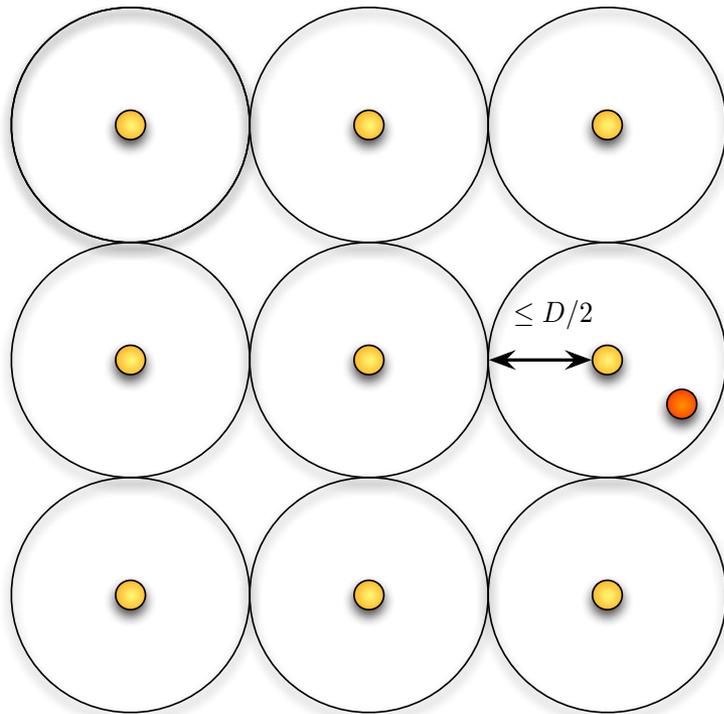
- **Spark(A)** = $\sigma(A)$ = min number of linearly dependent cols
- **Unique EXACT solution:** If $k < \sigma/2$, then there's a unique sparsest solution with sparsity k . [Donoho, Elad, 2003]
- **Coherence(A)** = $\mu(A)$ = max inner product between cols
- **Unique EXACT solution:** if $k < \mu/2$, then there's a unique sparsest solution with sparsity k . [Donoho, Elad, 2003]
- A = spikes and sines, $\mu(A) = 1/\sqrt{n}$
- Bounds = form of **Uncertainty Principle**



Unique sparse representations
!?

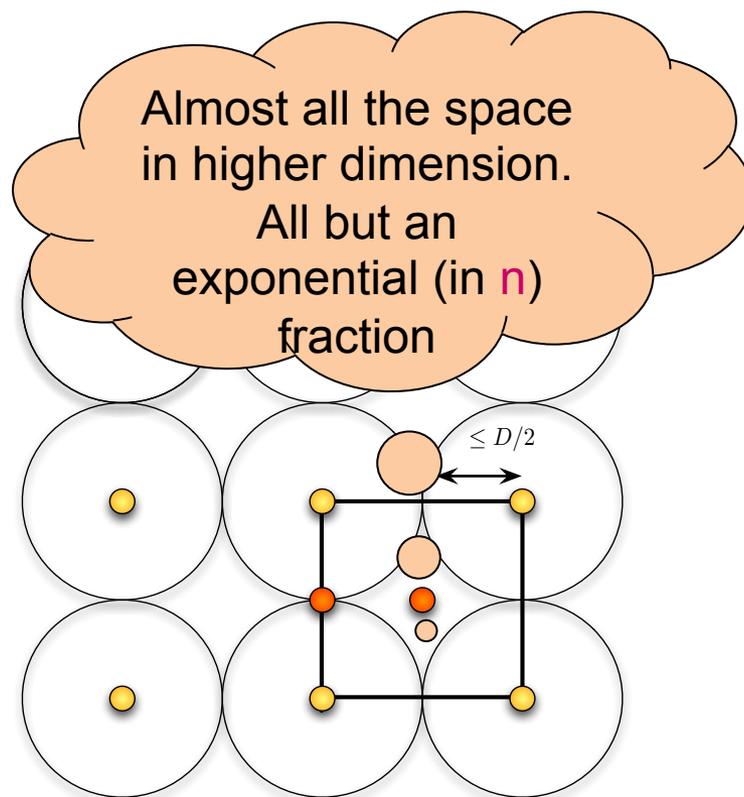
Error Correcting Codes: Unique decoding

- Distance of code = minimum distance between 2 codewords
- Receive corrupted codeword
- Return closest codeword
- Tolerate errors up to $D/2$



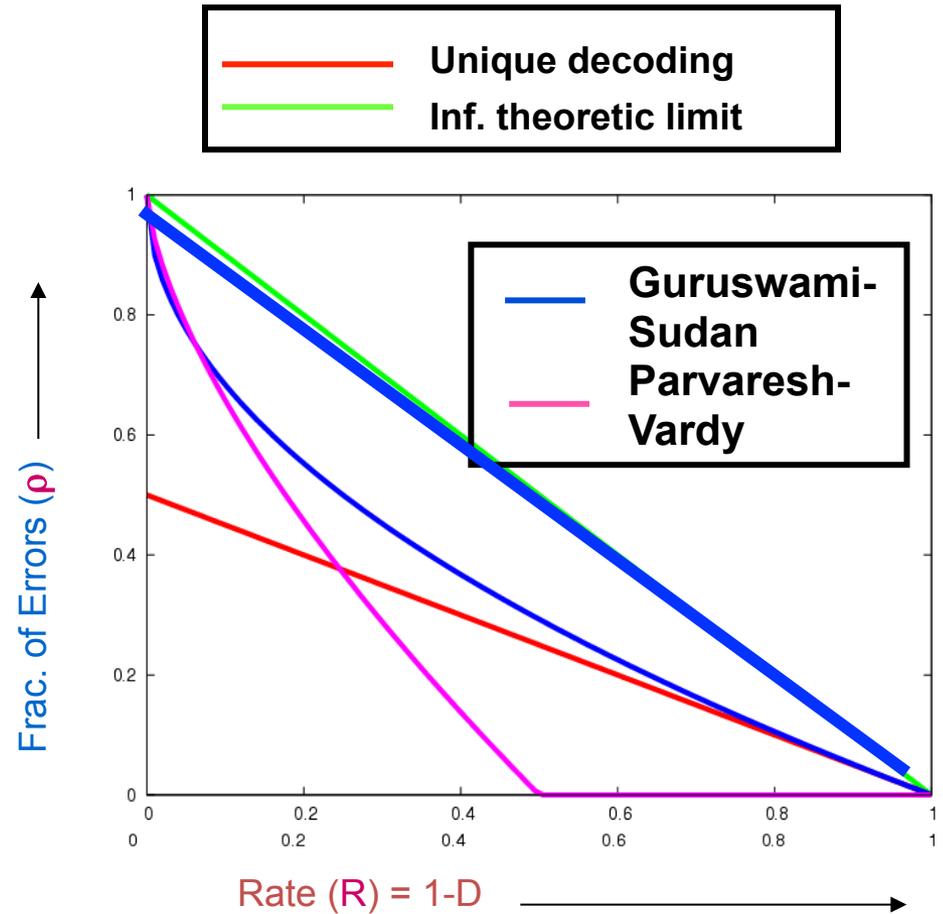
List Decoding ECC: definitions

- Return small list of codewords, with guarantee that transmitted codeword is in the list [Elias, 1957, Wozencraft, 1958]
- Formal defn: ρ = fraction of errors, list words differ from transmitted by no more than ρ



List Decoding ECC: implications

- Information Theory
 - Information theoretic limit $\rho < 1 - R$
 - Explicit constructions, efficient (?) algorithms: Folded RS codes
- Cryptography
 - Cryptanalysis of certain block-ciphers [Jakobsen, 1998]
 - Efficient traitor tracing scheme [Silverberg, Staddon, Walker 2003]
- Complexity Theory
 - Hardcore predicates from one way functions [Goldreich, Levin 1989; Impagliazzo 1997; Ta-Shama, Zuckerman 2001]
 - Worst-case vs. average-case hardness [Cai, Pavan, Sivakumar 1999; Goldreich, Ron, Sudan 1999; Sudan, Trevisan, Vadhan 1999; Impagliazzo, Jaiswal, Kabanets 2006]



List Decoding ECC: progression of results

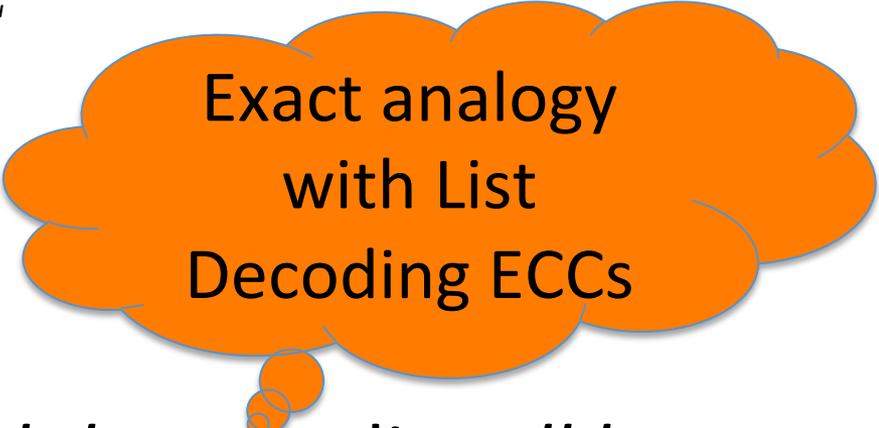
1. Combinatorial bounds on list size (Johnson bound)
2. Algorithms for finding list
3. Explicit ECCs that achieve bounds + practical algorithms

Sparse Approximation \leftrightarrow ECC

Sparse Approximation	ECC
Redundant dictionary	Codebook
Input signal	Received codeword + errors
Coherence	Distance
Redundancy	Rate
Spark	Spark
Best k-term approximation	Decoding
k=1	Closest codeword
k > 1	-----

List Sparse Approximation: Definitions

List SPARSE: Given A, b , and k , list *all* k -sparse x such that $\|Ax - b\|_2$ is minimized.



Exact analogy
with List
Decoding ECCs

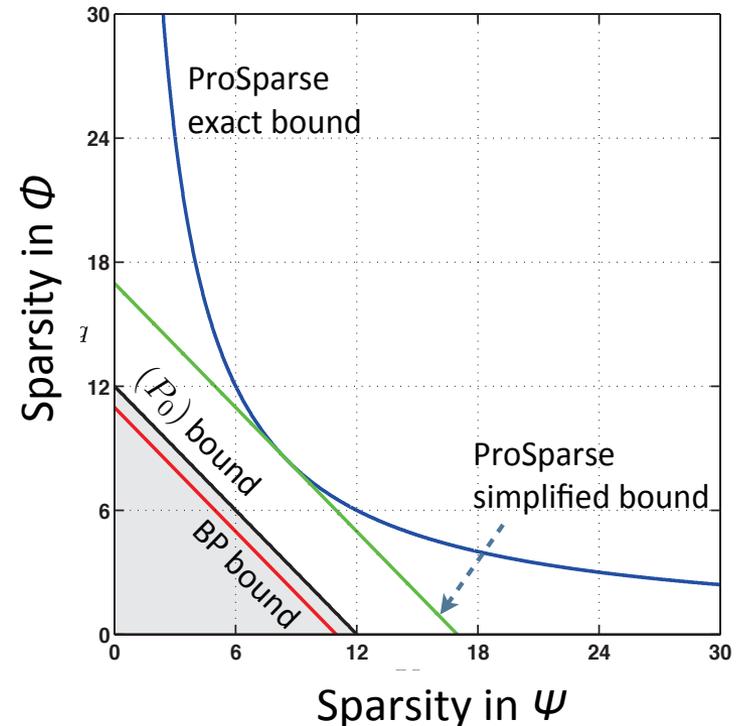
List APPROX: Given A, b, k , and ϵ , list *all* k -sparse x such that $\|Ax - b\|_2 \leq \epsilon$

List Sparse Approximation: Implications?

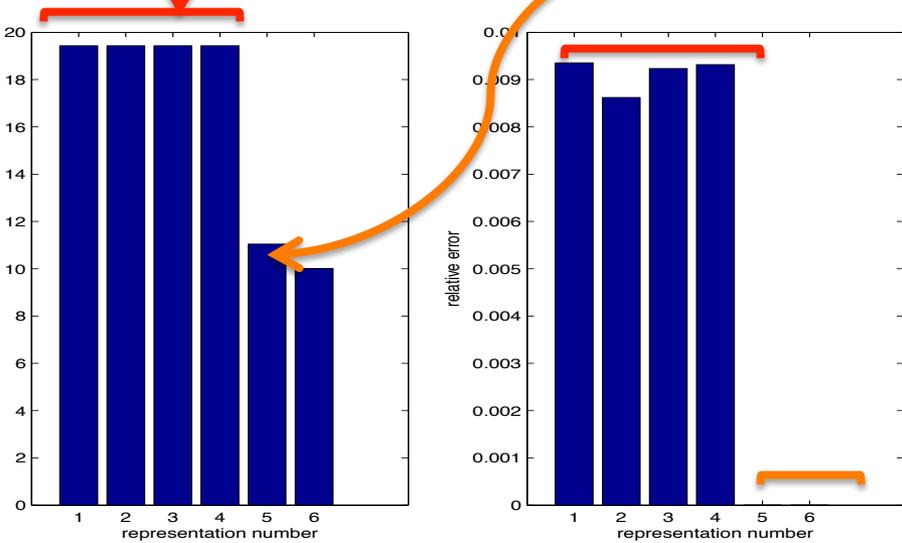
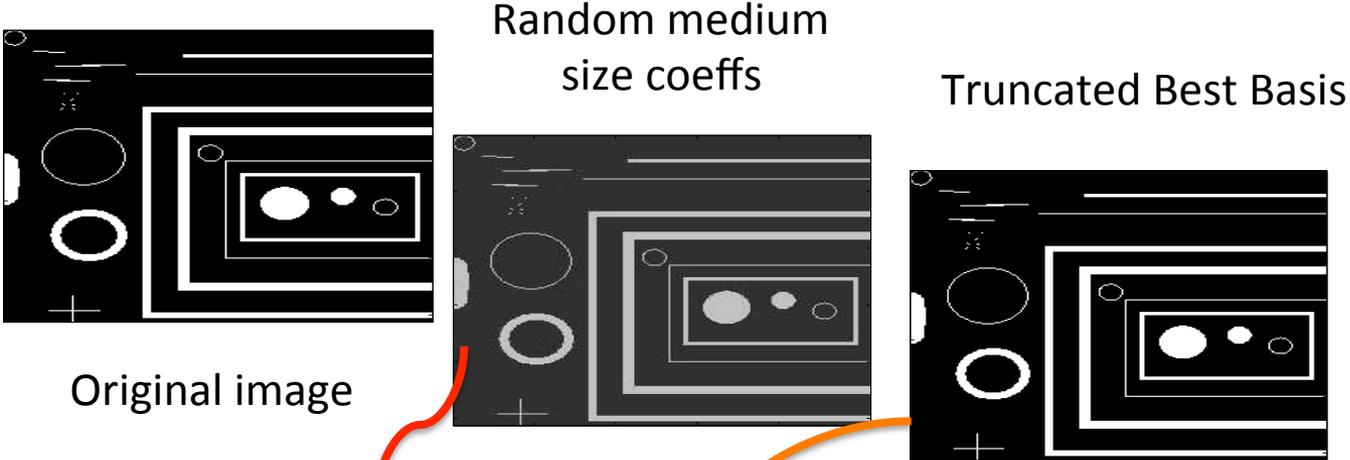
- **Algorithms' achievability**

[Dragotti, Lu 2013]

- $A = [\Psi, \Phi]$ union of ONBs (can be generalized)
- **ProSparse**: proto-list sparse approximation algorithm
- Returns list of *exact* representations “beyond” convex relaxation bound and unique repr. bound



List Sparse Approximation: Implications?



Equivalent error guarantees in compressed repr. versus ease of computation

Goals

1. Combinatorial bounds on list size
2. Dictionaries that achieve bounds
3. Practical algorithms + dictionaries

List Size: Clarifying definitions

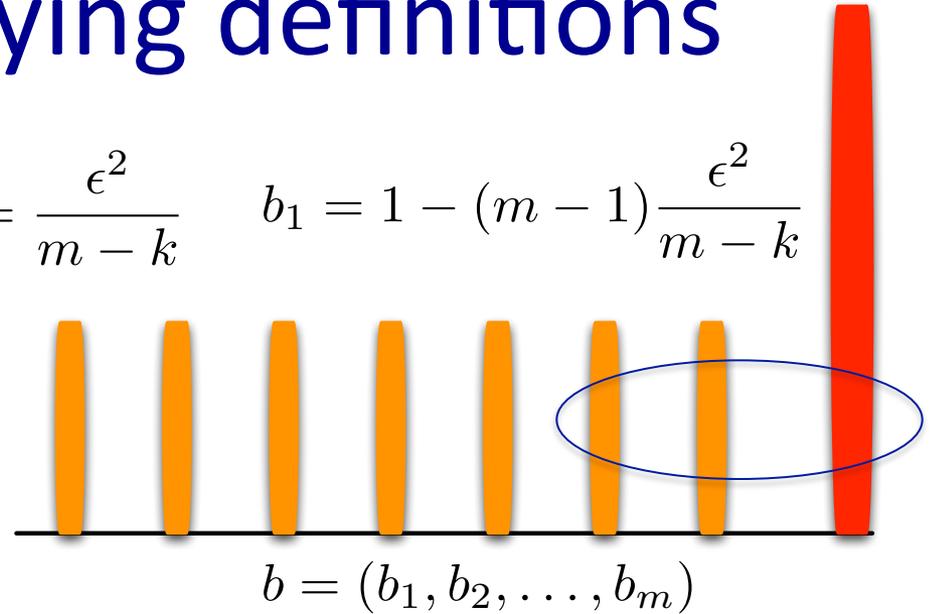
- $A = I_n$, $b = (1, 0, \dots, 0)$, $\epsilon = \sqrt{2}$
 - Choose $x_1 \in (1 - \sqrt{2}, 1 + \sqrt{2})$ $x_j = 0$
 - NOT a meaningful list!
- **List Approx:** Given A, b, k , and ϵ , $L(A, b, k, \epsilon)$ is number of distinct support sets of k -sparse solutions x such that $\|Ax - b\|_2 \leq \epsilon$

List Size: Clarifying definitions

- $A = I_n, \epsilon < \sqrt{\frac{m-k}{m}}$

$$b_i = \frac{\epsilon^2}{m-k} \quad b_1 = 1 - (m-1) \frac{\epsilon^2}{m-k}$$

coherence $\mu = 0$



- Any support set of size k that contains 1 has same error, $\binom{m-1}{k-1}$ different support sets
- $L(A, k, \epsilon, R)$ denotes the worst case bound on L over all b with the restriction that no atom appears in the support of more than R out of the L solutions.

List-Approx: Combinatorial bounds

- **Theorem:** (disjoint solutions)

If $\mu \leq \frac{1}{k}$

As long as the error $\varepsilon \leq 1 - \Omega(\mu k)$, the number of disjoint solutions is $O(1/(1 - \varepsilon^2))$.

- Extends Uncertainty Principle to more than two disjoint solutions and allows some approximation error.
- Unique decoding results as corollaries.

List-Approx: Combinatorial bounds

- **Theorem:**

Let $0 < \gamma < 1$. As long as

we have

If we consider only solutions where each atom appears only $o(L)$ times in the output list of size L , then L is bounded only by ε (and is independent of k and n).

$\overline{\gamma}$

$\gamma)$

b derived from vector that demonstrates tightness of UP

- **Lemma:** Let $A =$ Kerdock code dictionary, $\mu(A) = 1/\sqrt{n}$. For every $s < \sqrt{n}$, there is an input vector b s.t.
 - there are $\binom{n}{s}$ vectors x with sparsity $s\sqrt{n}$ and $Ax = b$
 - each atom appears in exactly s/n fraction of solns
- So,
 - For $\mu < 1/k$, $s = 1$, we have dictionary+input vector with $L(A, k, 0, 1) > n$.
 - For $s = \omega(1)$, $L(A, k, 0, o(L))$ can be super-poly in n .
- \rightarrow Coherence bound is tight!

Conclusions

1. Combinatorial bounds on list sizes ✓
2. Dictionaries that achieve bounds ✓
3. Practical algorithms + dictionaries ?