# Sub-linear Time Compressed Sensing and Group Testing via Left-and-Right-Regular Sparse-Graph Codes

Avinash Vem, Nagaraj T. Janakiraman, Krishna R. Narayanan ECE Dept., Texas A&M University

## Compressed Sensing(CS)

Recover a sparse signal  $\mathbf{x}$  from  $\mathbf{y}$ :  $\mathbf{y} = \mathbf{A}\mathbf{x} + \mathbf{w}$ 

- $\mathbf{x} N \times 1$  sparse signal
- $\mathbf{A} M \times N$  measurement matrix
- w -additive Gaussian noise
- $\mathbf{y} M \times 1$  measurement vector
- $|\{i: x_i \neq 0\}| = K$ . K << N

Metric of interest:

• Prob. of failure of support recovery  $\mathbb{P}_F := \Pr(\operatorname{supp}(\hat{\mathbf{x}}) \neq \operatorname{supp}(\mathbf{x}))$ 

# Group Testing(GT)

Recover a sparse signal  $\mathbf{x}$  from  $\mathbf{y}$ :  $\mathbf{y} = \mathbf{A} \odot \mathbf{x}$ 

- **x**, **y**, **A** are binary vectors/matrix respectively
- O: Matrix multiplication with "binary OR" instead of addition
- y-  $M \times 1$  measurement vector

Metric of interest:

• Prob. of miss:  $\mathbb{P}_m := \Pr(\hat{x}_i = 0, x_i = 1)$ 

## **Known Bounds for CS**

- For sub-linear sparsity, K = o(N),  $M = \Theta\left(K\log(\frac{N}{K})\right)$  is shown to be necessary and sufficient.
- In the linear sparse regime,  $K = \alpha N$ , it was shown that  $M = \Theta(N)$  measurements are sufficient.
- In [2], Li et al using left-regular sparse graph codes proposed a scheme that achieves  $O(K \log N)$  sample and decoding complexity

### Main Result (CS)

Sub-linear sparsity: For a given SNR, our scheme has

- Sample complexity of  $M = c_1 K \log(\frac{c_2 N}{K})$
- Decoding complexity of  $O(K \log(\frac{N}{K}))$
- $\mathbb{P}_{\mathrm{F}} \to 0$  asymptotically in K and N

Linear sparsity: Our scheme has

- Sample complexity of  $M = c_3 K \log K$
- Decoding complexity of  $O\left(K\log(\frac{N}{K})\right)$
- $\mathbb{P}_{\mathrm{F}} \to 0$  asymptotically in K and N

## Bounds for Group Testing

- We assume all the  $\binom{N}{K}$  K-sparse sets are equi-probable
- At least  $\log_2 \binom{N}{K}$  tests are necessary
- For large K and N,  $\log_2 \binom{N}{K} \approx K \log_2 (\frac{N}{K})$
- In [1], Lee et al using left-regular sparse graph codes proposed a scheme that achieves  $cK \log N$  testing complexity

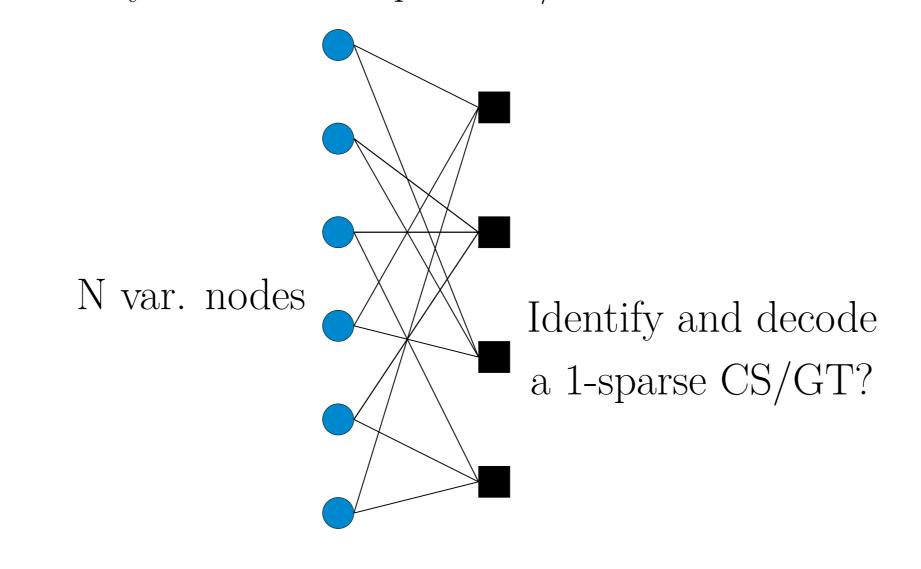
## Main Result (GT)

**Sub-linear sparsity:** Let  $K \in o(N^{\frac{p}{p+1}})$ , for some  $p \in \mathbb{Z}$ 

- Number of tests  $M = 2(p+1)c(\epsilon) K \log_2(\frac{c_1 N}{K})$
- Decoding complexity of  $O\left(K\log(\frac{N}{K})\right)$
- $\mathbb{P}_m \leq \epsilon$  w.h.p. asymptotically in K and N
- e.g. for  $\epsilon = 10^{-5}$ ,  $c(\epsilon) = 9.63$

## Idea: Divide-and-Conquer

- $\bullet$  Original problem is K-sparse CS/GT
- ullet Divide N nodes into non-disjoint bins
- Can you solve for 1-sparse CS/GT at a bin?



K-sparse CS/GT

# Divide: $(\ell, r)$ Bipartite Graph

- N Variable(left) nodes:  $x_i$ . Each node has (left) degree:  $\ell$
- Bin(right) nodes: Choose  $M_1 = cK$  bins (sub problems)
- Each bin has (right) degree r. Gives  $r = \frac{N\ell}{cK}$
- ullet Connections between  $N\ell$  edges on each side are random

# Conquer: 1-sparse CS

ullet At each bin, use code words of an error control code  ${\cal C}$ :

$$\mathbf{y}_i = x_{i1}\mathbf{c}_1 + x_{i2}\mathbf{c}_2 + \dots x_{ir}\mathbf{c}_r + \mathbf{w}_i$$

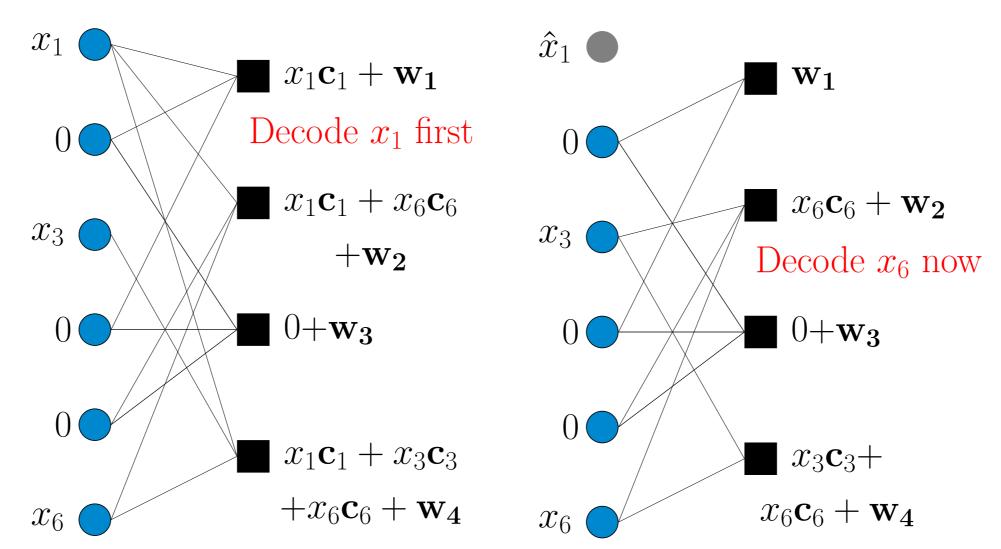
• If it is 1-sparse i.e. only one  $x_{ik} \neq 0$ : (call it singleton):

$$\mathbf{y}_i = x_{ik}\mathbf{c}_k + \mathbf{w}_i$$

- A channel coding problem if  $sign(x_{ik})$  is known
- From channel coding,  $\dim(\mathbf{c}_j) > \frac{\log r}{C_{\mathrm{Sh}}} \approx \Theta(\log \frac{N}{K})$
- After decoding index k, need to decode the value of  $x_{ik}$
- Let  $x \in \mathcal{X} := \{\pm A : A \in \mathcal{A}\}$  be a discrete and finite set
- We decode  $\hat{x}_{ik} \in \mathcal{X}$  to be the value that best fits  $\mathbf{y}_i \mathbf{c}_k^{\dagger} / ||\mathbf{c}_k||^2$

## Reconstruction via Peeling

- Assume we can conquer the 1-sparse CS/GT at a bin
- If a singleton bin is found, decode the index and the value
- Peel off the decoded variable node value from other bins



- Continue peeling iteratively until no new singletons are found
- $\exists$  threshold  $c_*$  such that for  $M_1 > c_*K$  bins, peeling decoder recovers all nodes w.h.p asymptotically

## Conquer: 1 and 2-sparse GT

- Let  $\mathbf{b}_k$  be the binary expansion of k;  $\overline{\mathbf{b}}_k$  it's complement
- At each bin, test result vector would be:

$$\mathbf{y}_i = x_{i1} \begin{bmatrix} \mathbf{b}_1 \\ \overline{\mathbf{b}}_1 \end{bmatrix} \vee x_{i2} \begin{bmatrix} \mathbf{b}_2 \\ \overline{\mathbf{b}}_2 \end{bmatrix} \vee \ldots \vee x_{ir} \begin{bmatrix} \mathbf{b}_r \\ \overline{\mathbf{b}}_r \end{bmatrix}$$

• If it is 1-sparse i.e. only one  $x_{ij} = 1$ , trivial to decode j:

$$\mathbf{y}_i = egin{bmatrix} \mathbf{b}_k \ \mathbf{b}_k \end{bmatrix}$$

- Non-linear OR operation poses problem in peeling; can't remove a decoded variable node from connected bins
- If a bin is 2-sparse i.e.  $x_{ij}, x_{ik} = 1$  and an index j is known: refer to as resolvable double-ton

$$\mathbf{y}_i = egin{bmatrix} \mathbf{b}_j \ \mathbf{ar{b}}_j \end{bmatrix} ee egin{bmatrix} \mathbf{b}_k \ \mathbf{ar{b}}_k \end{bmatrix}$$

- bins with more than 2 non-zero variables (multi-ton) are unusable due to OR
- For  $M_1 > c(\epsilon)K$  bins, just using singletons and double-tons, peeling like iterative decoder recovers  $(1 \epsilon)$  fraction of non-zero nodes w.h.p asymptotically

### Numerical Results

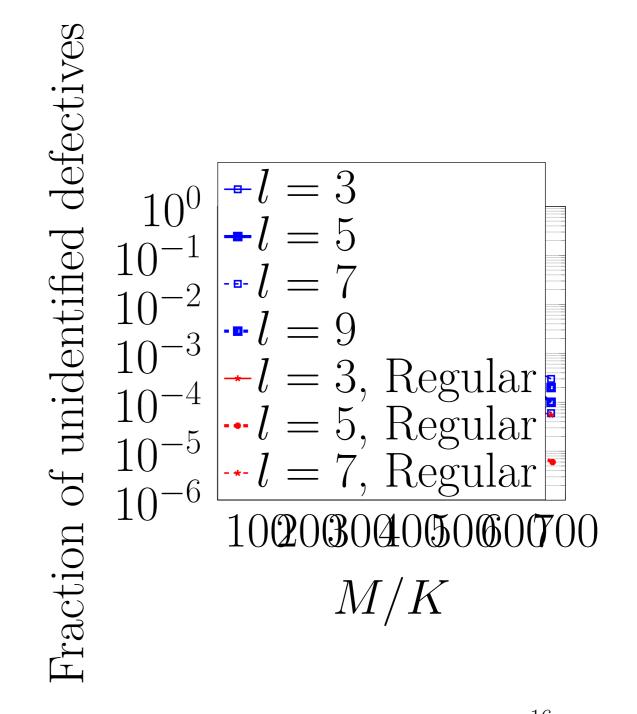


Figure: Monte Carlo simulations for  $K=100, N=2^{16}$ . For a given  $\ell$  the bin detection size is fixed and we vary the number of bins. The plots in blue indicate the scheme in [1] and the plots in red indicate our scheme based on left-and-right-regular bipartite graphs for various left degrees  $\ell \in \{3,5,7\}$ .

#### Conclusions

Compressed Sensing: We propose a scheme that has

- order optimal sample complexity of  $O(K \log(\frac{N}{K}))$
- sub-linear optimal decoding complexity:  $O(K \log(\frac{N}{K}))$

Group testing: We propose a scheme that achieves

- order optimal testing complexity:  $O(K \log(\frac{N}{K}))$
- sub-linear optimal decoding complexity:  $O(K \log(\frac{N}{K}))$

#### References

- [1] K. Lee, R. Pedarsani, and K. Ramchandran, "Saffron: A fast, efficient, and robust framework for group testing based on sparse-graph codes", arXiv preprint, arXiv:1508.04485, 2015.
- 2] X. Li, S. Pawar, and K. Ramchandran, "Sub-linear time compressed sensing using sparse-graph codes", in *Proc. Int. Symp. Inform. Theory*, pp. 1645-1649, 2015.
- [3] A. Vem, N. Janakiraman, and K. Narayanan, "Sub-linear time compressed sensing for support recovery using left and right regular sparse-graph codes", in *Proc. Inform. Theory. Workshop*, pp. 429-433, 2016.