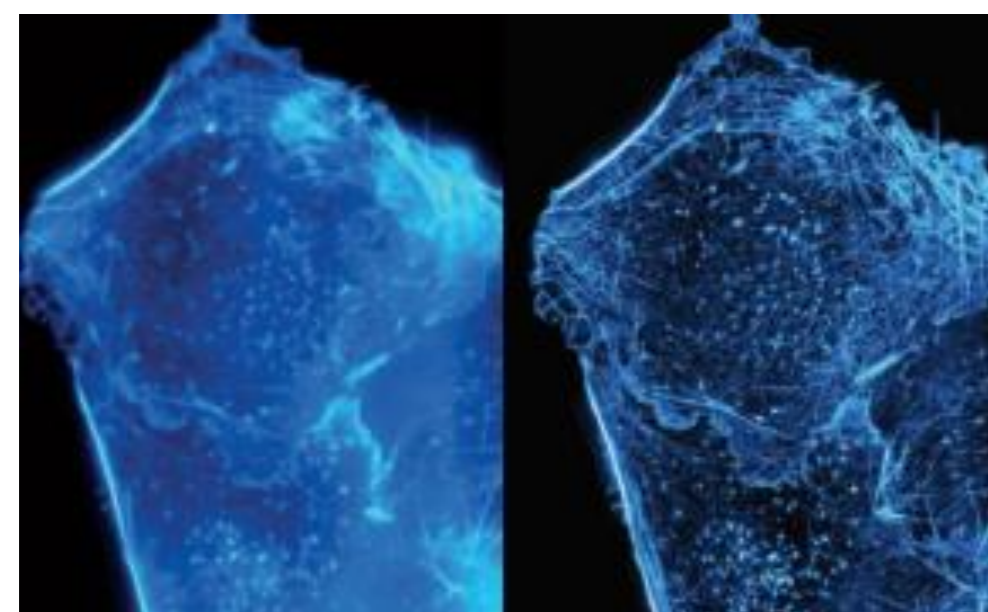
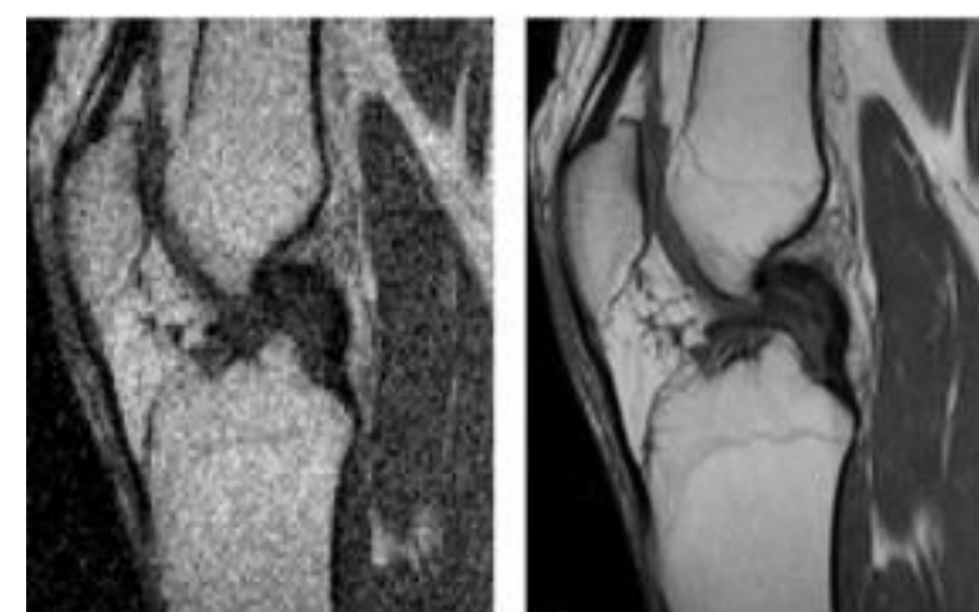




Application: Super-resolution



Microscopy imaging [EKB]



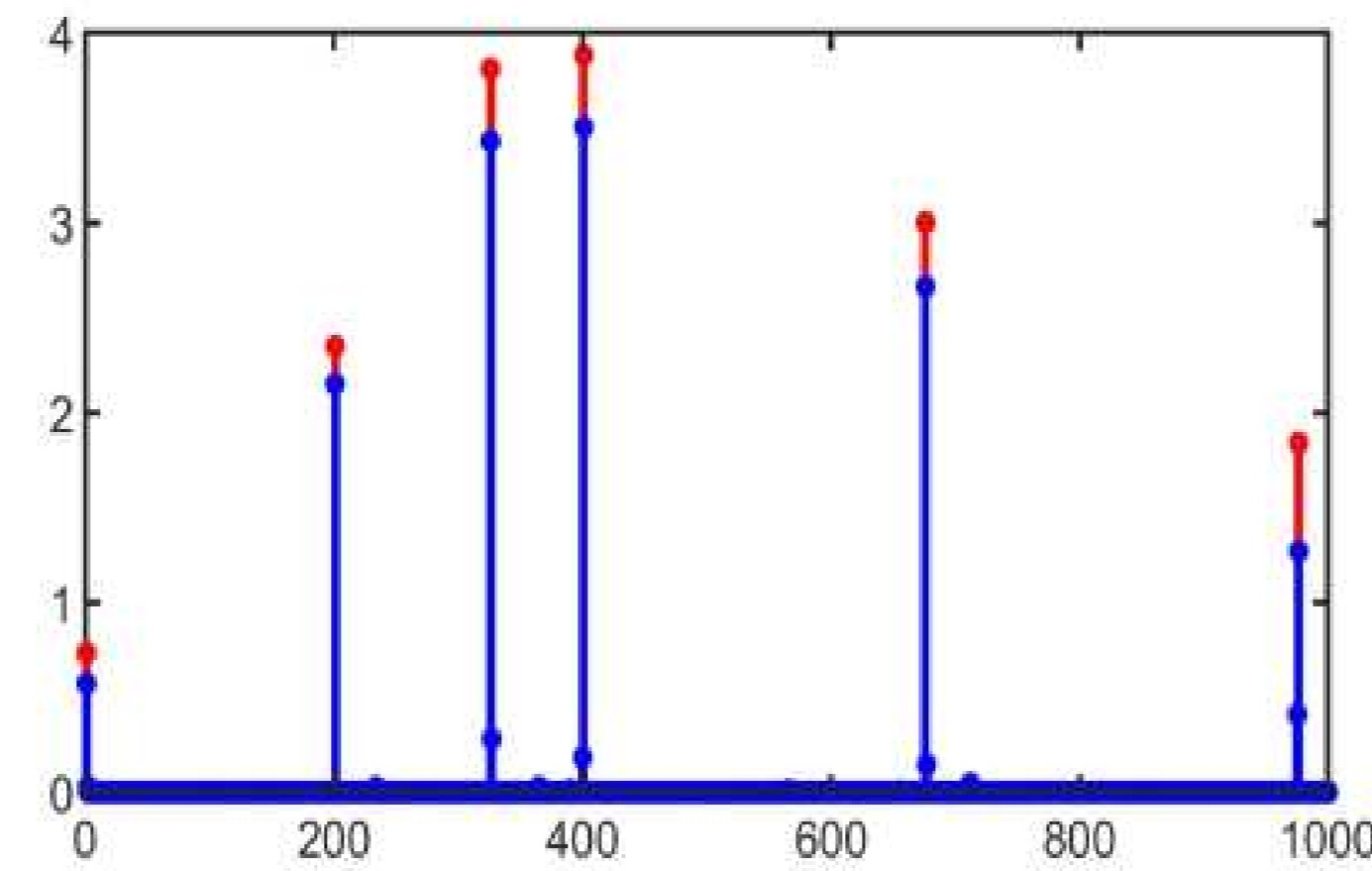
Medical imaging [Medium]



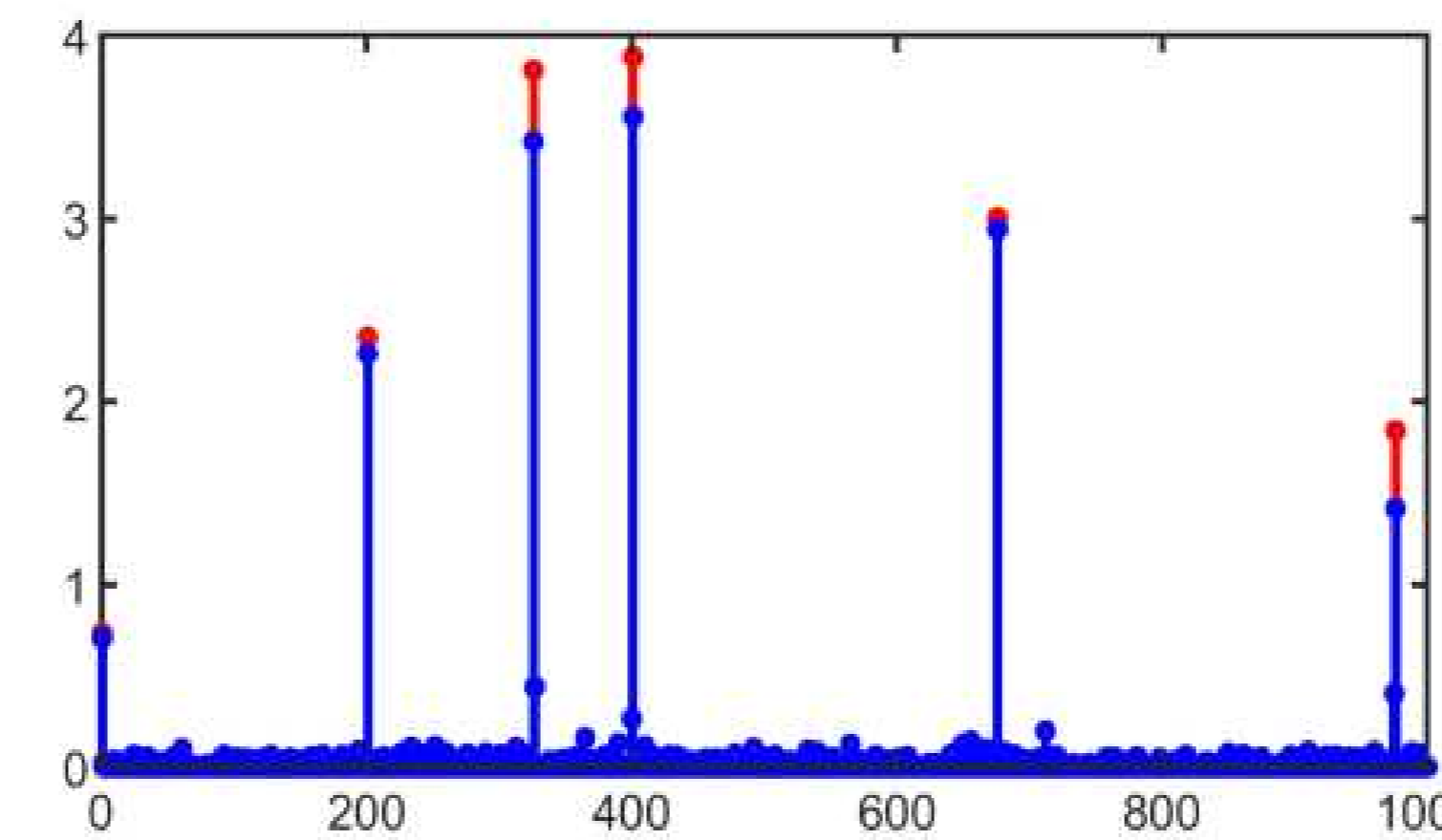
Astronomy [Nap]

Goal: Reconstruct the sparse target signal from its low-frequency measurements

Motivation: Comparison between Non-overfitting and Overfitting Recoveries



Non-overfitting BPDN-2 Recovery



Overfitting BPDN Recovery

Key observation: Estimates of non-overfitting and overfitting recoveries are similar.

Question: Is the prior knowledge of the noise necessary for a stable estimate?

Proof Sketch

Quotient property

Lemma 1: $Au = w, \|u\|_q \leq c(M, N)\|w\|$

Null space property

Lemma 2: $\|V_S\|_q \leq \frac{\gamma}{s^{1-1/q}} \|V_{S^c}\|_1 + \tau \|Av\|$

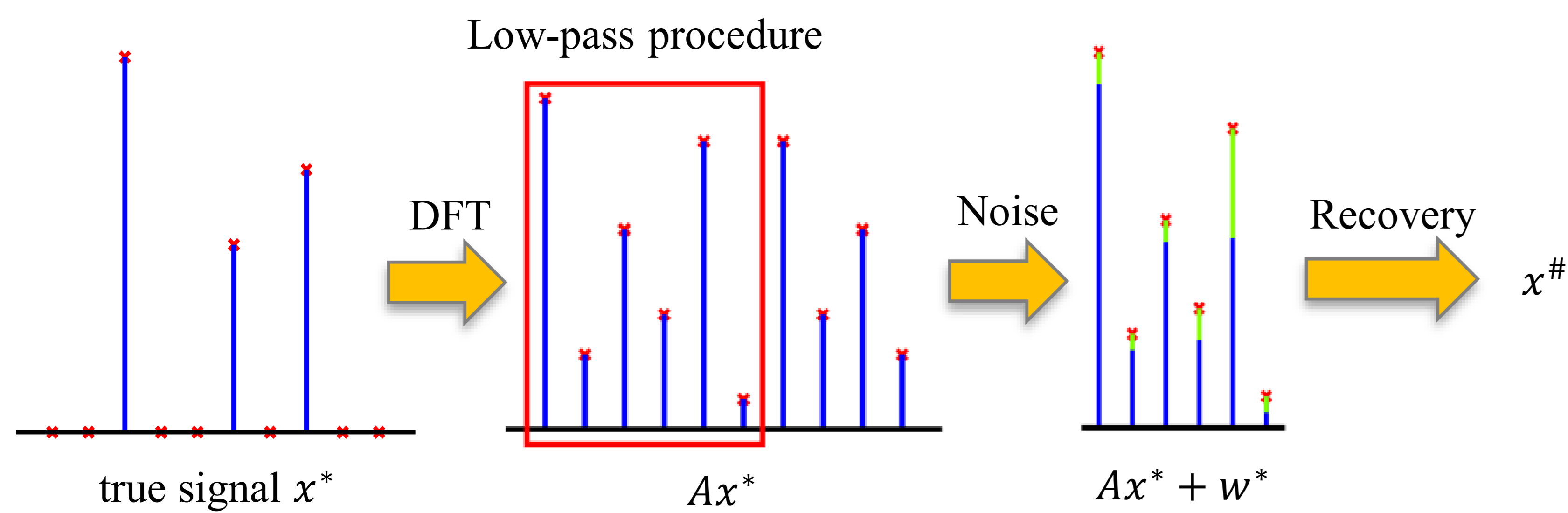
Overfitting stability guarantee

Challenge: RIP/Coherence based techniques don't apply to deterministic A

New insight:

An interpolation-based technique under the separation condition replaces NSP.
 $\|V_S\|_1 \leq (1 - \rho)\|V_{S^c}\|_1 + \|q\|_2 \|Av\|_2$

Super-resolution Recovery Framework



Studied overfitting recovery:

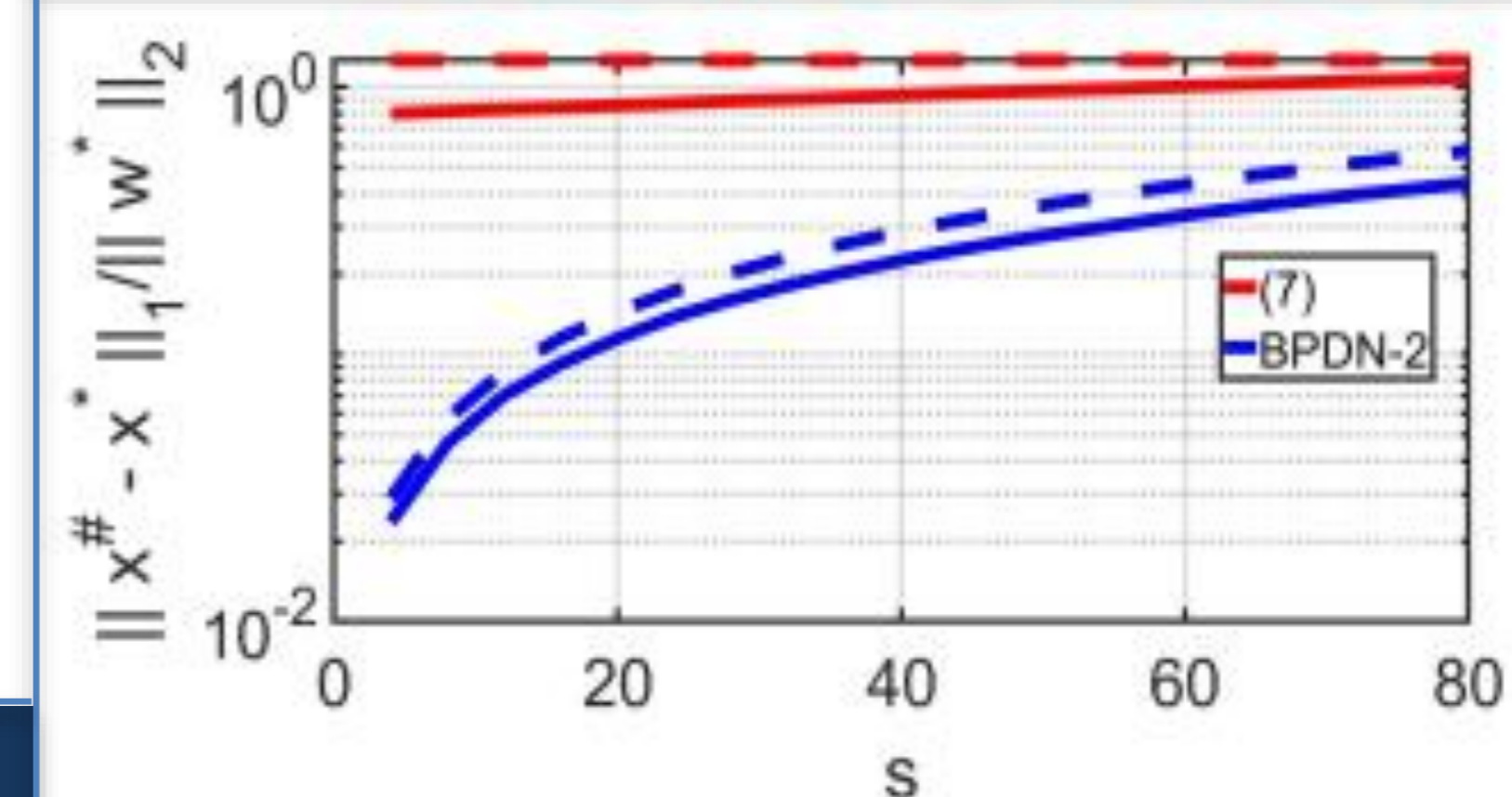
$$x^\# = \min_x \|x\|_1, \text{ s.t. } Y = Ax$$

No prior knowledge of noise needed

Simulations

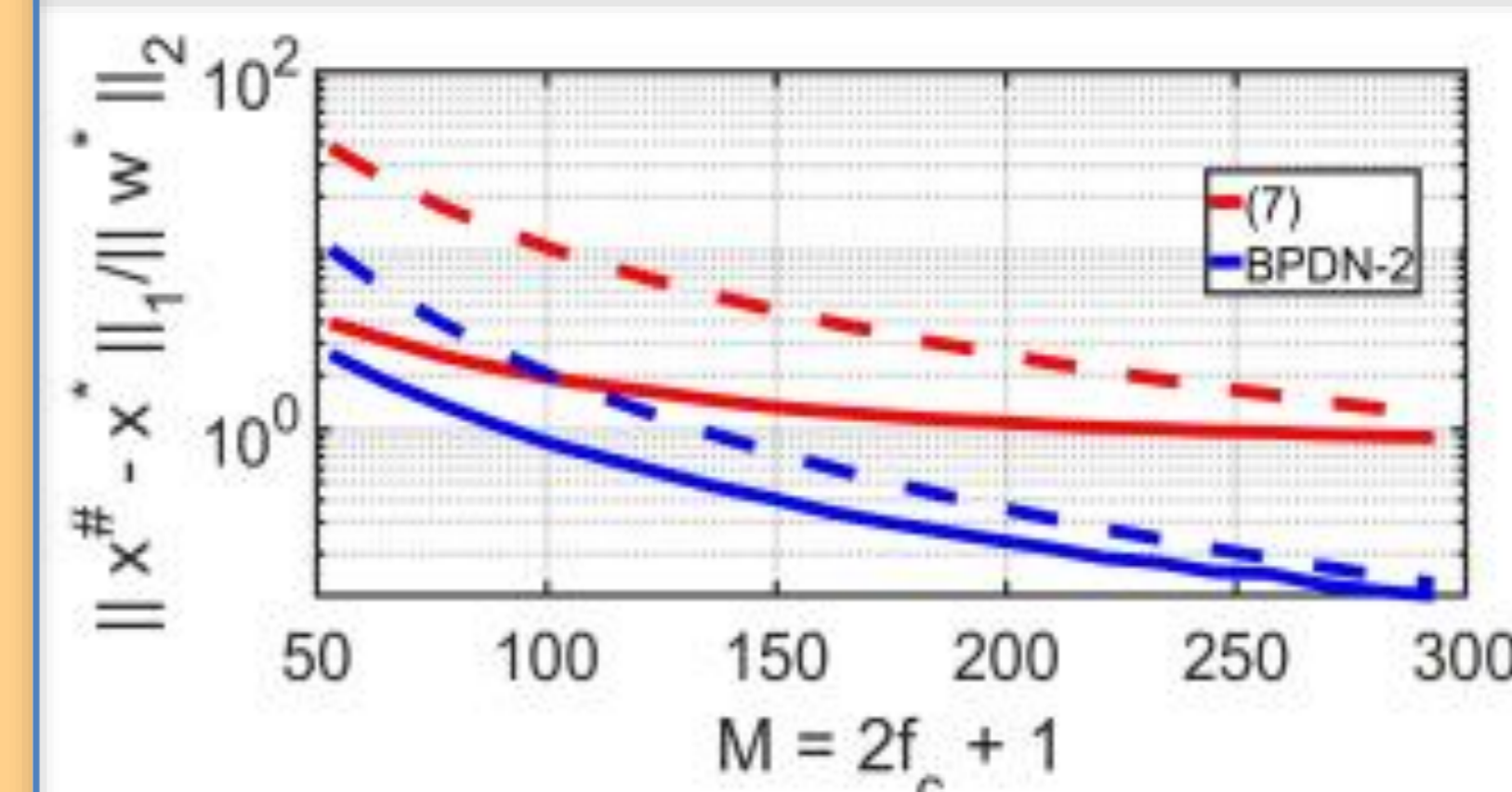
Comparison of the estimation error $\|x_{BP}^\# - x^*\|_1$.
Setting: $N = 1600, \sigma_x^2 = 10, \sigma_w^2 = 1$ and $\epsilon = 1.02 \|w^*\|_2$.

Experiment A



Fix $M = 521$, change s

Experiment B



Fix $s = 8$, change M

Main Result

Theorem: consider the overfitting recovery algorithm with partial DFT matrix A and the true sparse source signal x^* . Suppose the true support S satisfies the separation condition,

$$\min_{m, n \in S, m < n} \left\{ \frac{|m - n|}{N}, \frac{|m - n + N|}{N} \right\} \geq \frac{2.5}{f_c} \text{ with } f_c \geq 128,$$

then for any additive noise w^* , we have

$$\|x_{BP}^\# - x^*\|_1 \leq \frac{4}{\rho} \|w^*\|_2 \leq 12 \|w^*\|_2 \left(\frac{N}{f_c} \right)^2.$$