

### Compressive Sensing (CS)

#### The CS Framework

Random projections for measuring  $y = \Phi x$

$$M \times 1 \begin{matrix} y \\ \vdots \\ y \end{matrix} = \begin{matrix} \Phi \\ \vdots \\ \Phi \end{matrix} \begin{matrix} x \\ \vdots \\ x \end{matrix}$$

$M \times N$        $N \times 1$        $K$

$$M = O(K \log(N/K))$$

Reconstruct data  $x$  from measurements  $y$  using **convex optimization**

#### Wideband Multichannel Signals

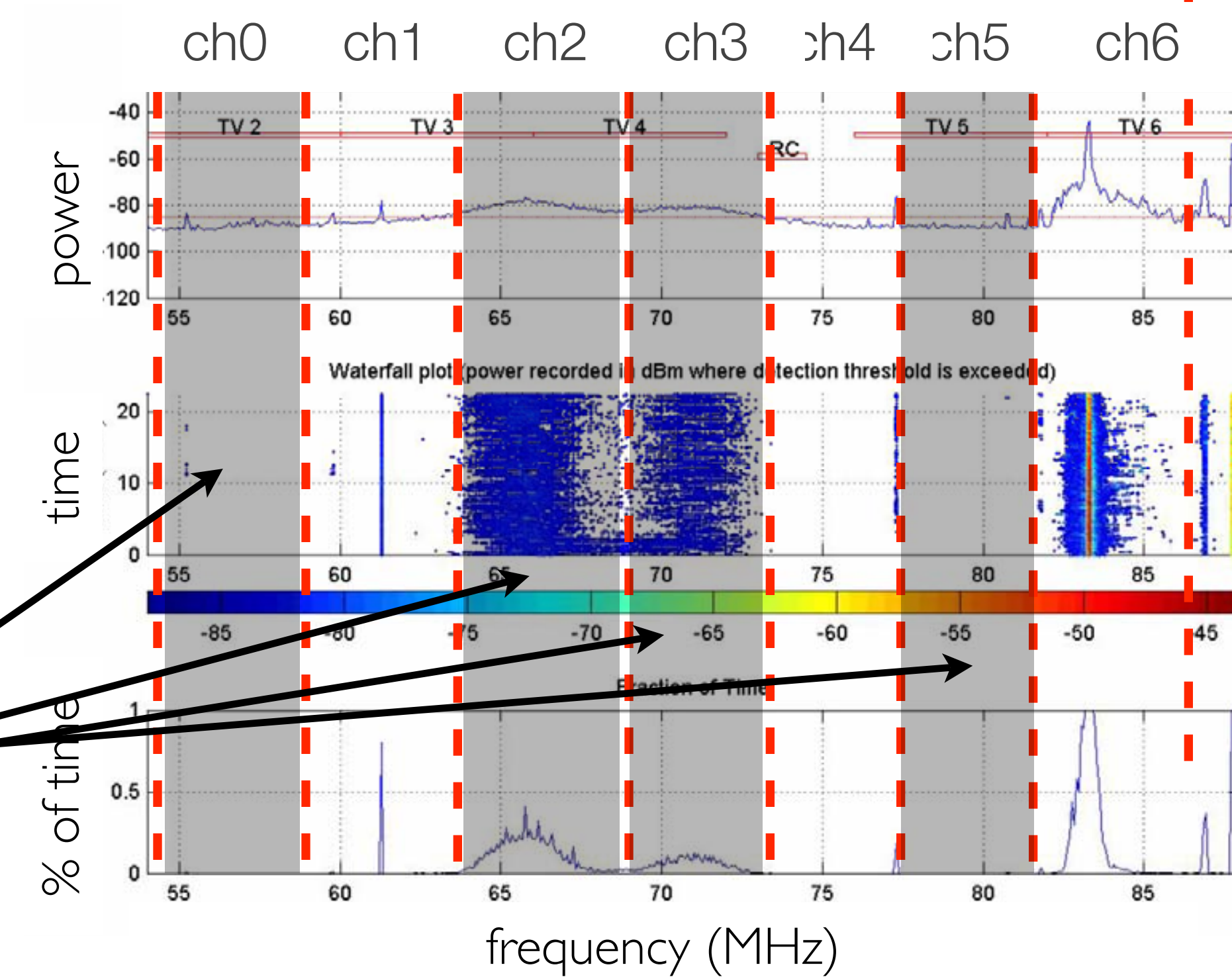
Goal Acquire signals from multiple (potentially discontinuous) channels

- channels may be known *a priori*
- channels must be sparse when concatenated

#### Example

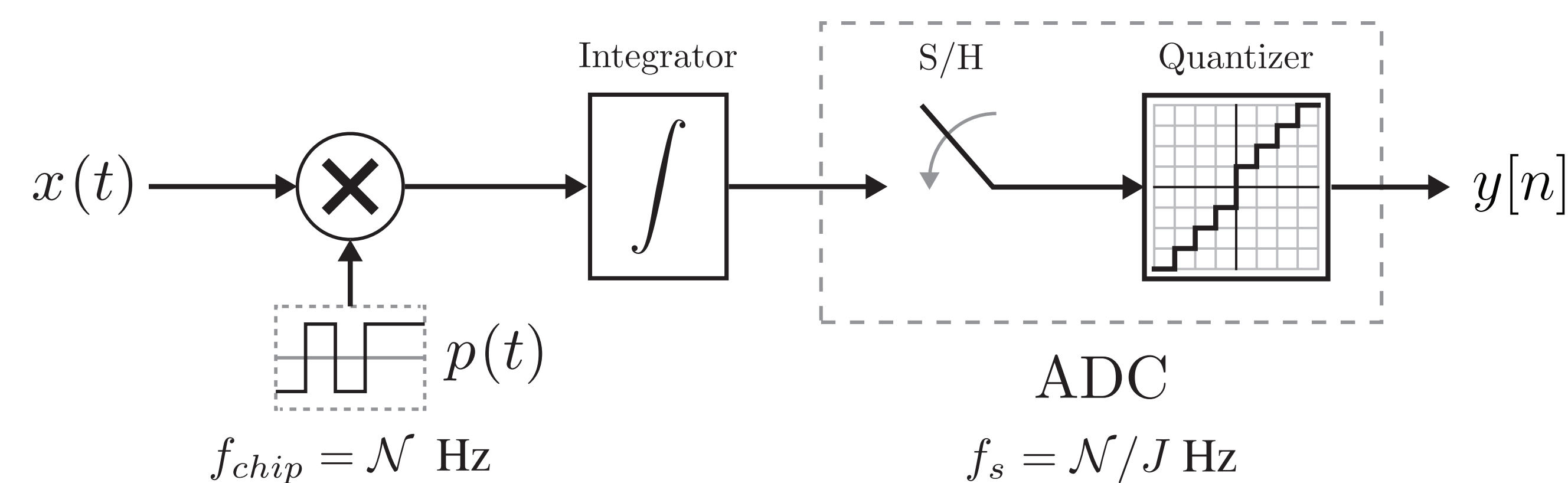
wireless spectrum  
most of spectrum not occupied

some channels not occupied



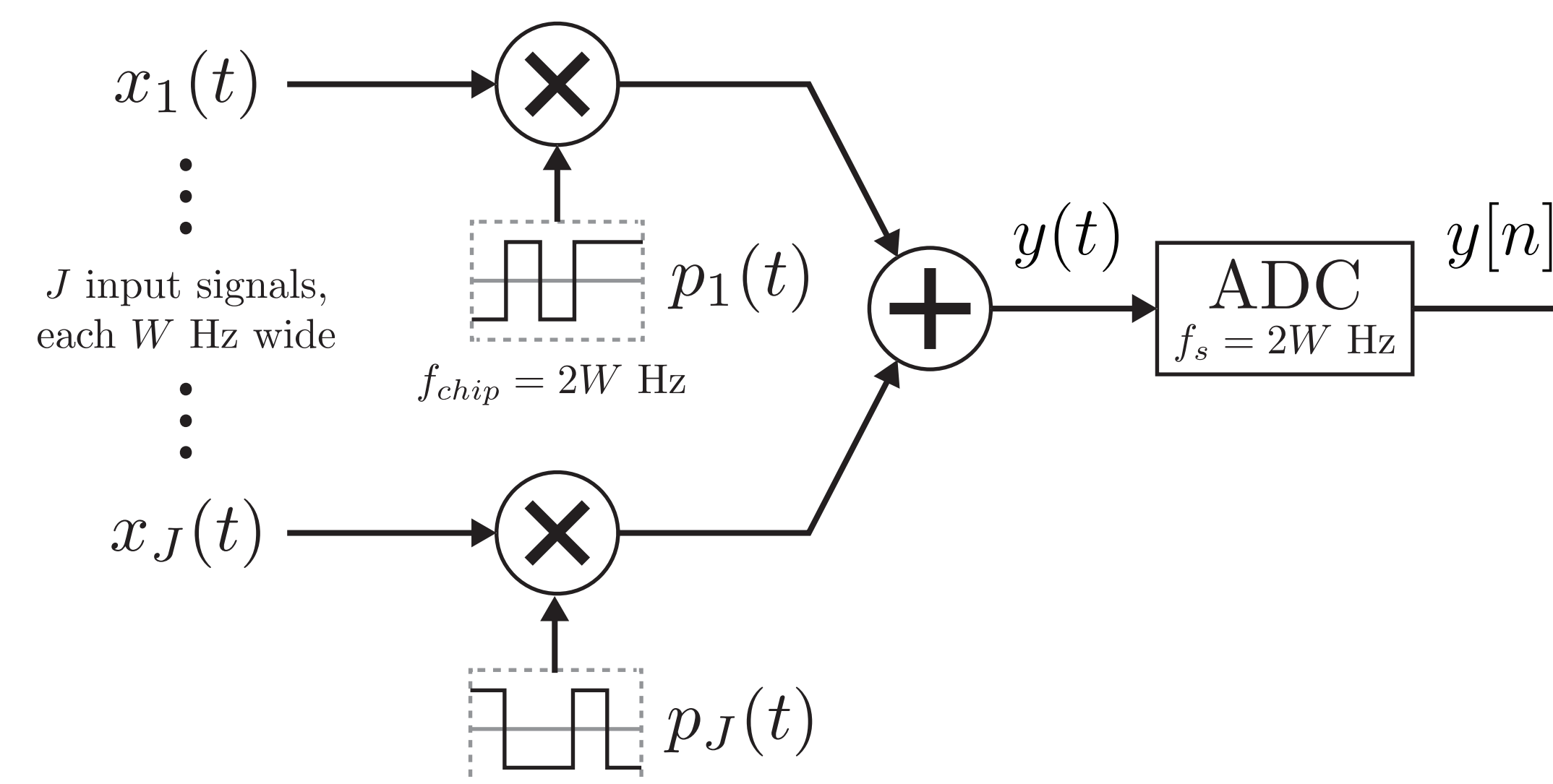
only interested in these channels

#### Random Demodulator (RD)

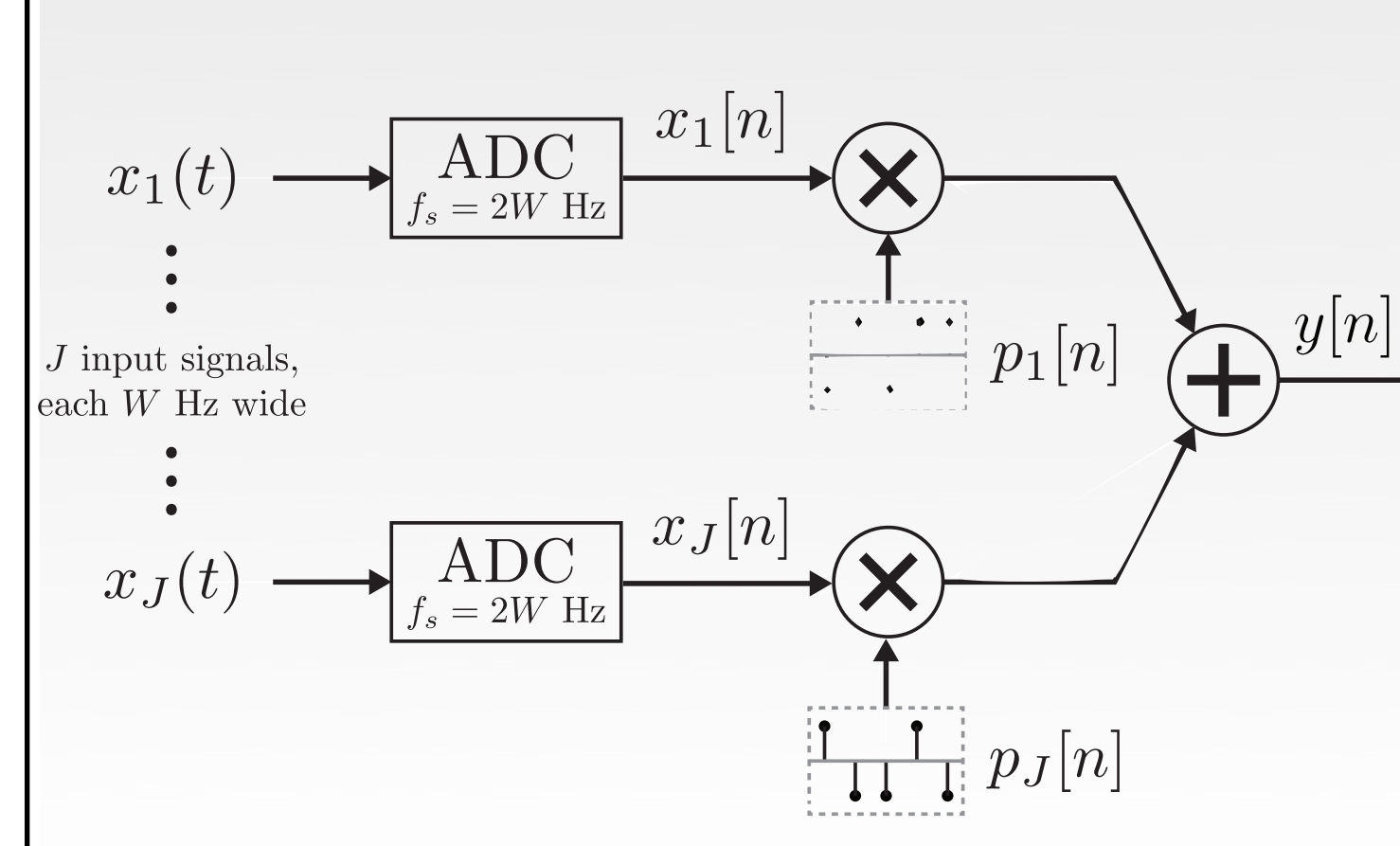


### The Compressive Multiplexer

#### System Model



#### Equivalent System



#### CS Model

$$\Phi = \begin{bmatrix} -1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 0 & -1 & 0 \\ 0 & 0 & -1 & 0 & 0 & -1 & 0 & 0 & -1 \end{bmatrix}$$

$\Phi_1$        $\Phi_2$        $\Phi_3$

$$A = [\Phi_1 \mathcal{F}, \Phi_2 \mathcal{F}, \dots, \Phi_J \mathcal{F}]$$

recover  $\alpha$  from  $y = A\alpha$  / DFT matrix

#### Theory

Theorem Fix  $\delta \in (0, 1)$  There exists  $C_0$  such that when

$$W \geq C_1 K \log^4(JW)$$

then  $A$  satisfies the RIP with probability  $1 - C_0^2 / \delta C_1^2$  [Neelamani, Romberg]

#### Algorithm Pairings

##### Trivial Reconstruction

$$\hat{x}_i = \Phi_i y$$

$$= \Phi_i \left( \sum_j \Phi_j x_j \right)$$

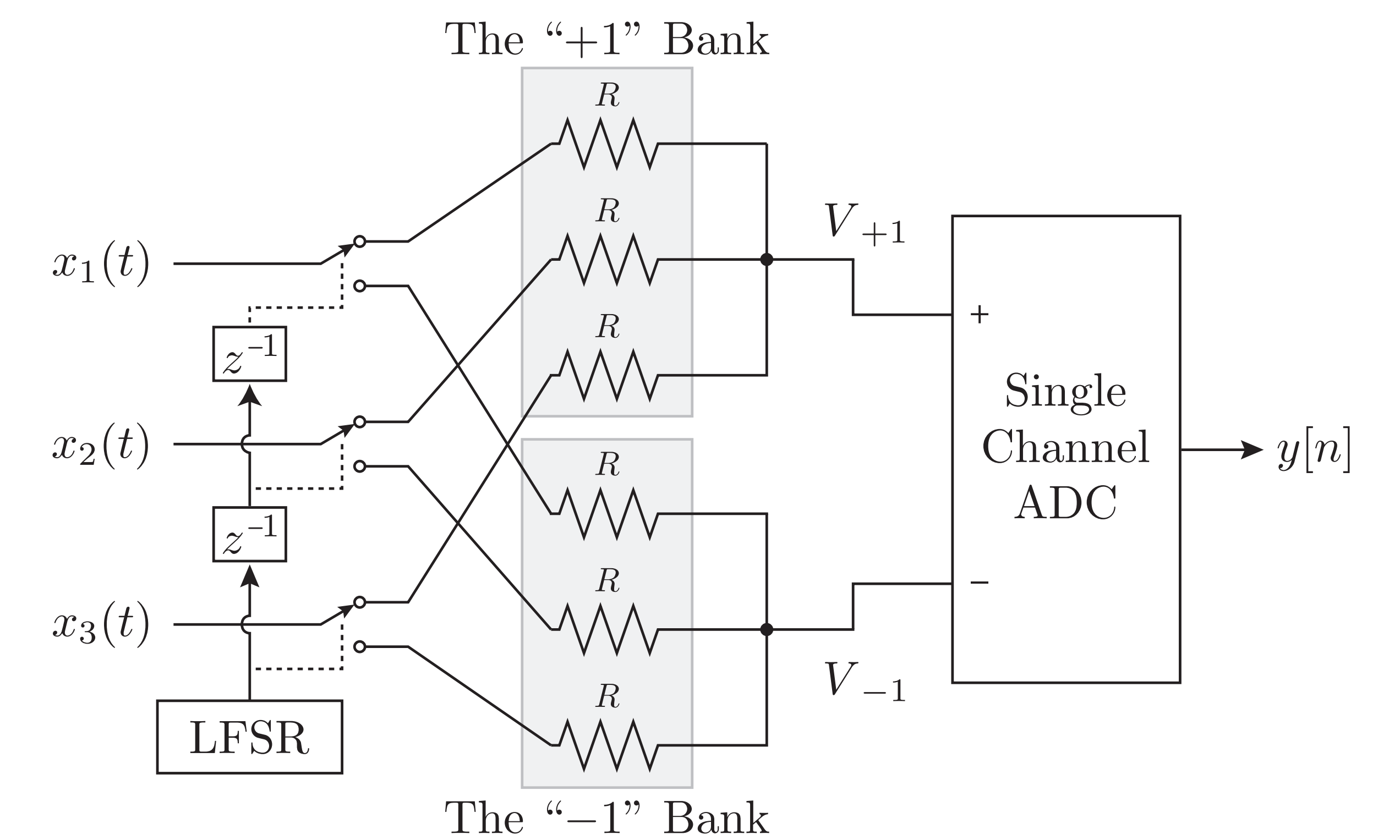
$$= \underbrace{x_i}_{\text{channel } i \text{ signal}} + \underbrace{\sum_{i \neq j} \Phi_i \Phi_j x_j}_{\text{incoherent noise}}$$

##### Block Coordinate Relaxation (BCR)

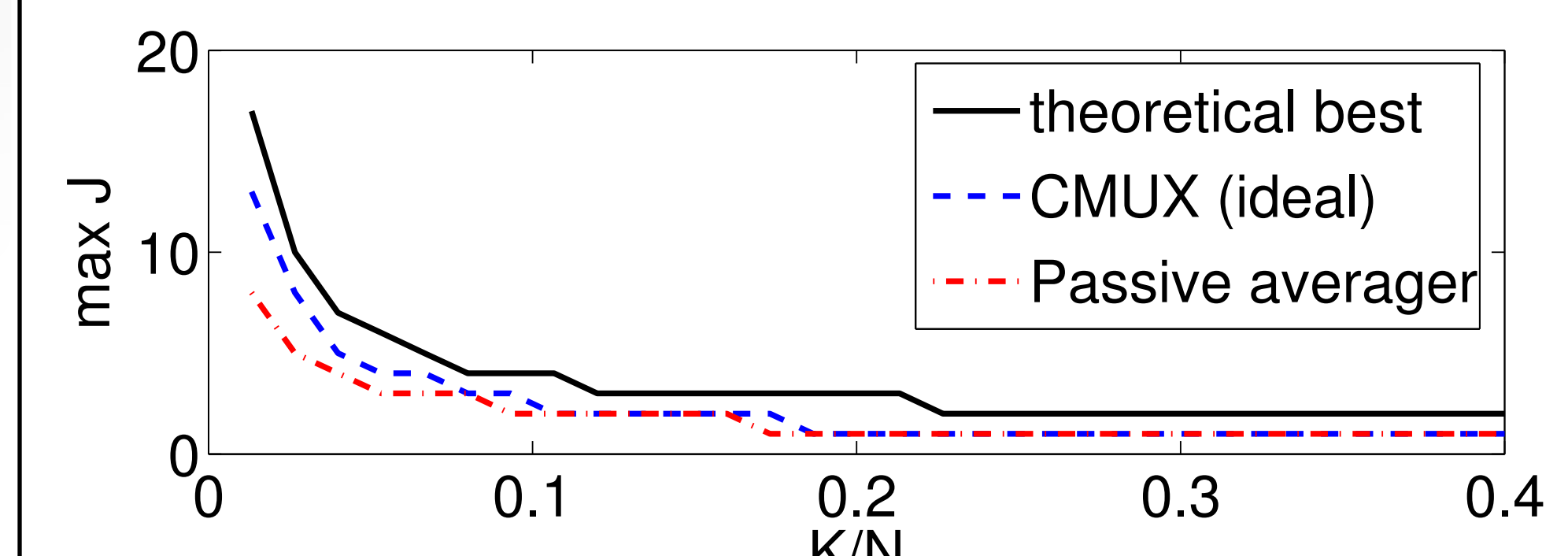
- i) choose new channel index  $j \in \{1, \dots, J\}$
- ii) subtract contribution from current estimate (except channel  $j$ )  
 $\mathbf{r} = \mathbf{y} - A_{\setminus j} \alpha_{\setminus j}$
- iii) update channel coefficients  
 $\alpha_j = \mathcal{S}(A_j^T \mathbf{r})$      $\mathcal{S}(z) = \frac{z}{|z|} (|z| - \lambda)_+$

### Practice

#### The Passive Averager



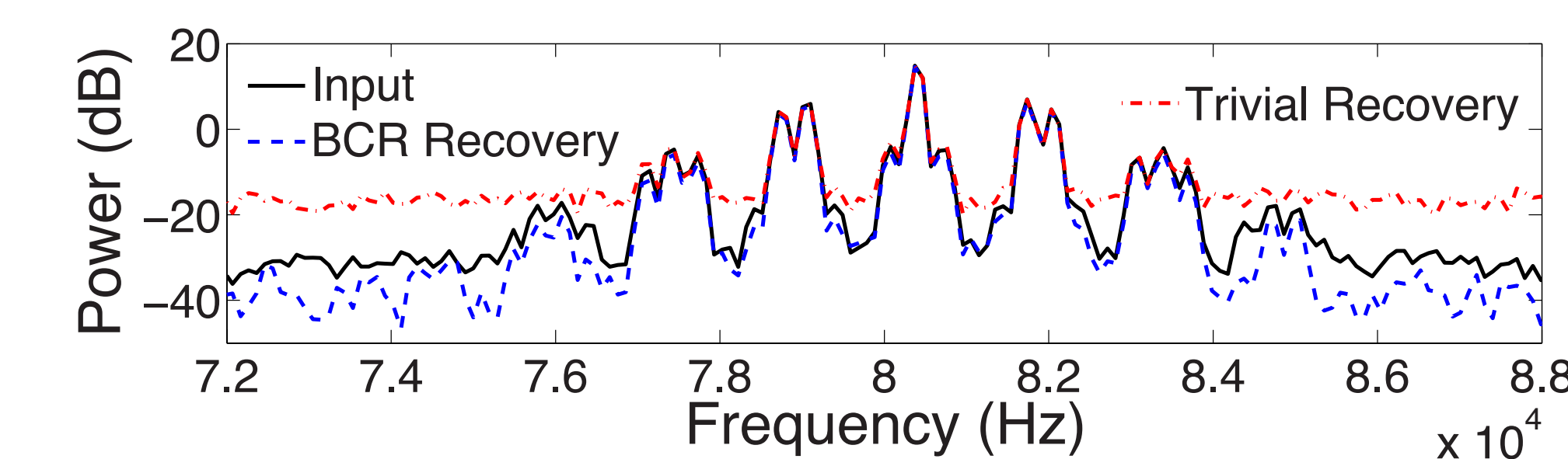
#### Numerical Simulations



$N = JW = 5000$   
Search for best  $J$ , given  $K, N$   
90% recovery from 1000 trials

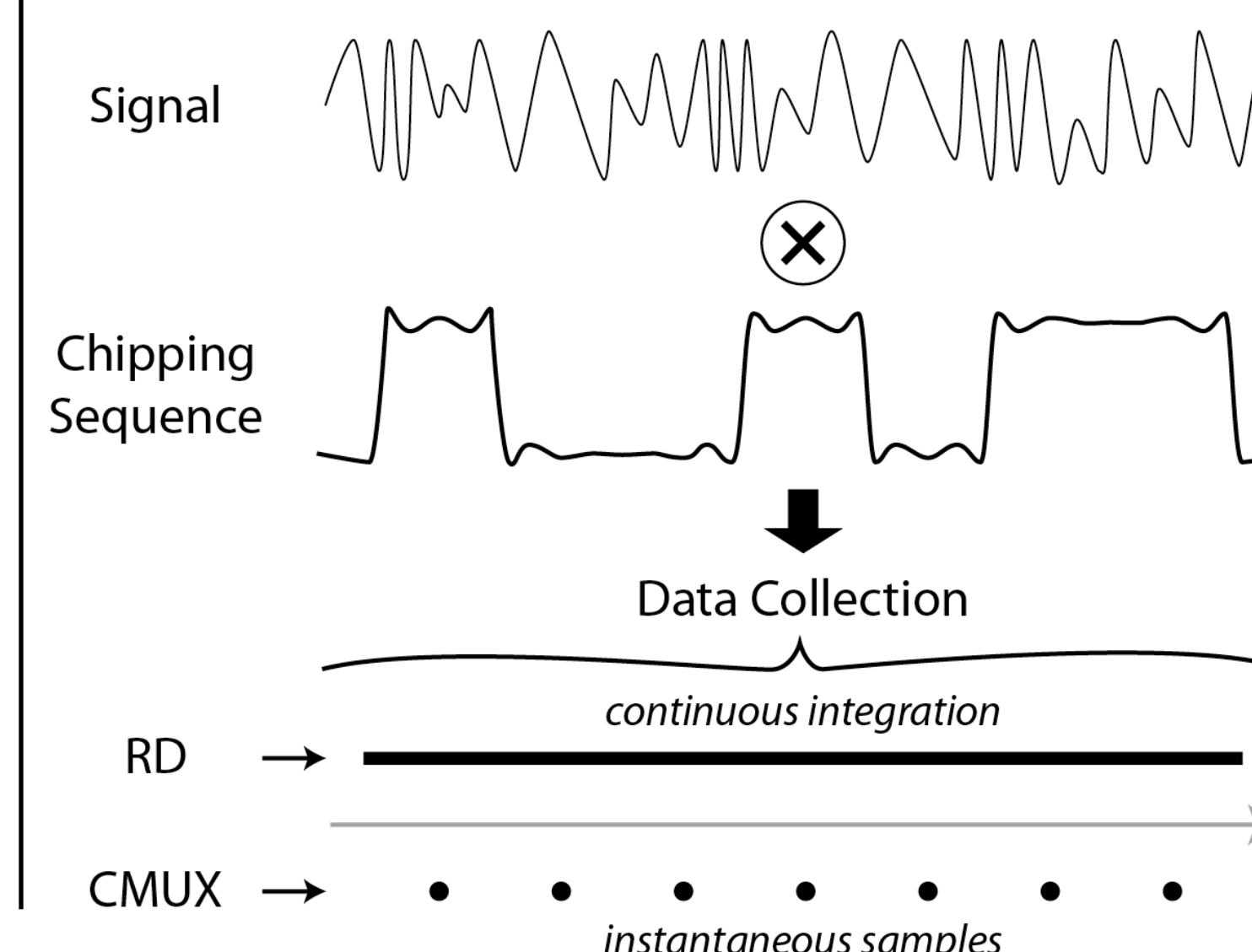
2 FM voice signals  
approximately 12 KHz wide (30dB SNR)

each signal in a different 400 KHz channel



5 channels total

#### Tradeoffs



CMUX does not integrate variations in chipping sequence waveforms

CMUX is easier to calibrate than the RD (no need to model filter)

CMUX subsampling rate is limited by number of channels