

A Low-Complexity Receiver for CP-OFDM in Doubly-Selective Channels

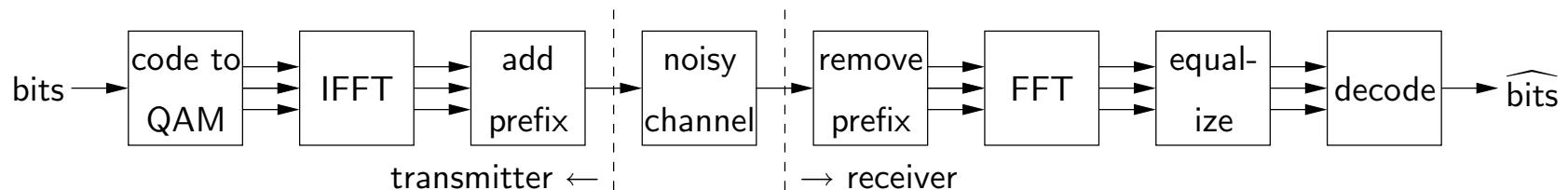
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Cyclic Prefix OFDM (CP-OFDM):

- Uses FFT for efficient modulation/demodulation.
- $\mathcal{O}(\log N)$ operations/symbol for FFT length N .
- Perfect interference suppression when
 - cyclic prefix length > channel delay spread,
 - channel is time-invariant over the FFT-block duration.

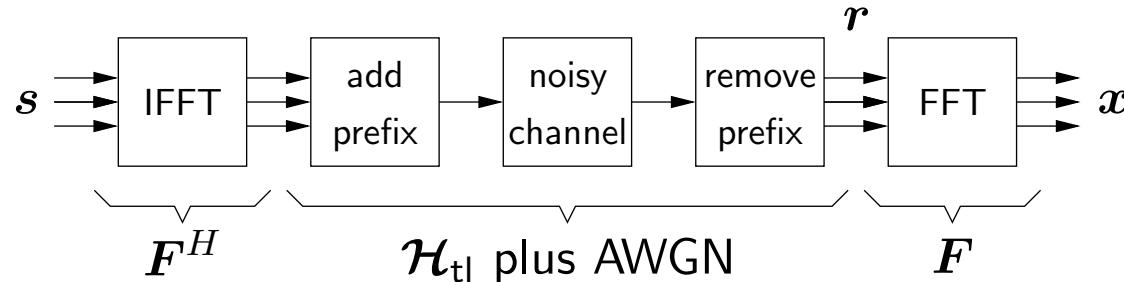


Doubly-Selective Channel:

- Long OFDM symbol motivated by reduction of cyclic-prefix redundancy.
- Channel time-variation present in applications with high mobility and/or high carrier frequency.

~> *Need a way to handle significant channel time-variation within an OFDM symbol interval.*

System Model:



$$\mathbf{r} = \mathcal{H}_{\text{tl}} \mathbf{F}^H \mathbf{s} + \boldsymbol{\nu}$$

$$\mathbf{x} = \underbrace{\mathbf{F} \mathcal{H}_{\text{tl}} \mathbf{F}^H}_{\mathcal{H}_{\text{df}}} \mathbf{s} + \underbrace{\mathbf{F} \boldsymbol{\nu}}_{\mathbf{w}}$$

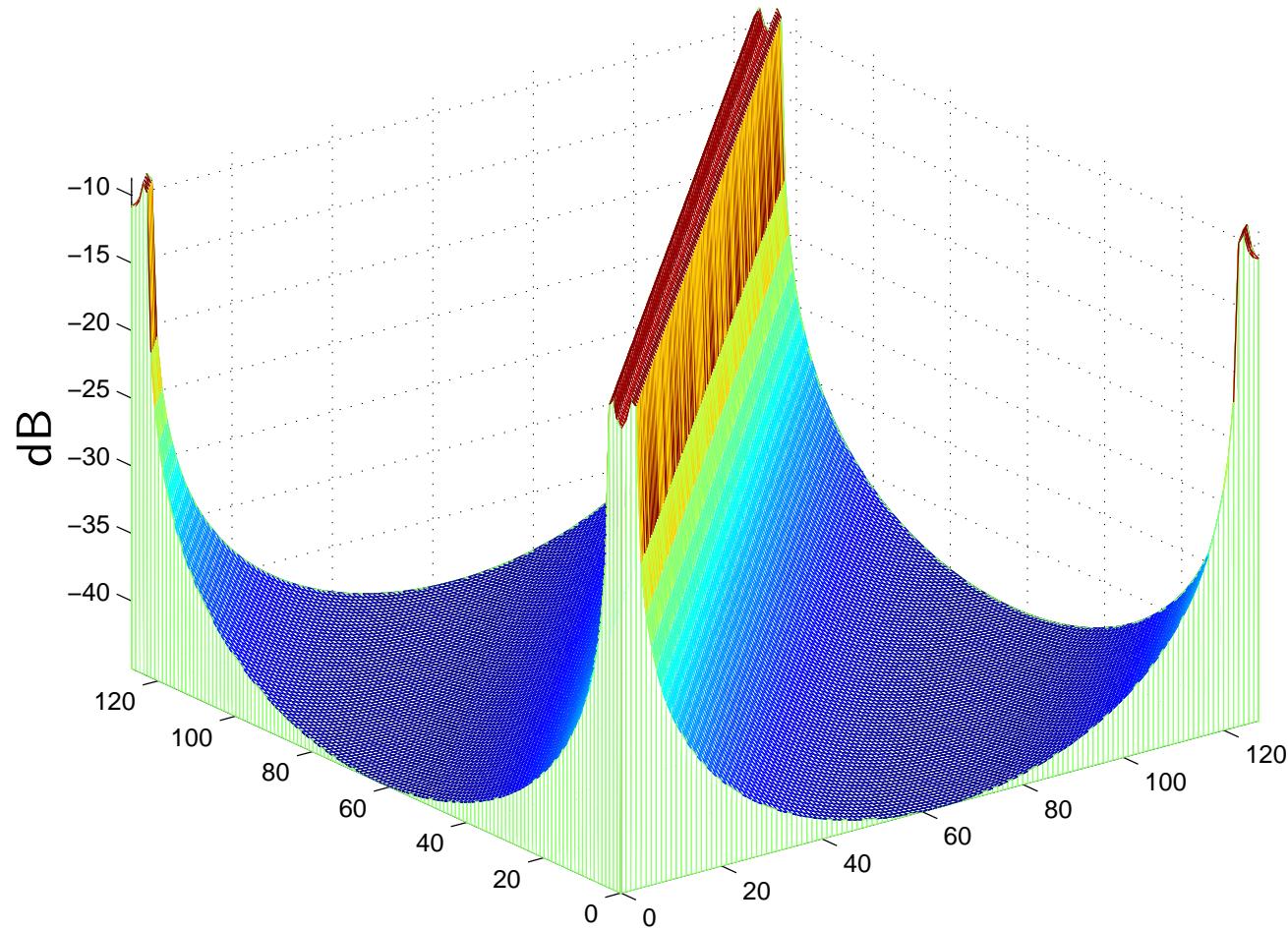
where

\mathcal{H}_{tl} = circular-convolution matrix

\mathcal{H}_{df} = sub-carrier coupling matrix

\mathcal{H}_{df} diagonal iff channel is LTI (assuming adequate prefix length)

Rayleigh $E\{|\mathcal{H}_{df}]_{m,n}|^2\}$ for $N = 128$ and $f_d = 0.03$:



Subcarrier Coupling Matrix \mathcal{H}_{df} :

$$\mathcal{H}_{\text{df}} = \begin{pmatrix} h_{\text{df}}(0, 0) & h_{\text{df}}(-1, 1) & \dots & h_{\text{df}}(1-N, N-1) \\ h_{\text{df}}(1, 0) & h_{\text{df}}(0, 1) & \dots & h_{\text{df}}(2-N, N-1) \\ \vdots & \vdots & \ddots & \vdots \\ h_{\text{df}}(N-1, 0) & h_{\text{df}}(N-2, 1) & \dots & h_{\text{df}}(0, N-1) \end{pmatrix}$$

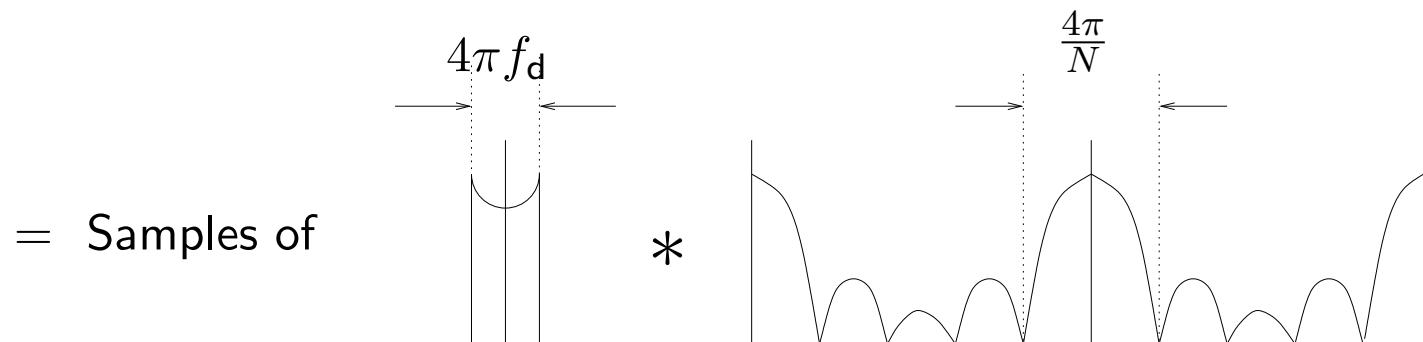
$$h_{\text{df}}(\nu, k) := \frac{1}{N} \sum_{n=0}^{N-1} \sum_{l=0}^{N-1} h_{\text{tl}}(n, l) e^{-j \frac{2\pi}{N} n\nu} e^{-j \frac{2\pi}{N} lk}$$

= response at carrier $k+\nu$ to an impulse applied at carrier k

$$h_{\text{tl}}(n, l) := \text{response at time } n \text{ to an impulse applied at time } n - l$$

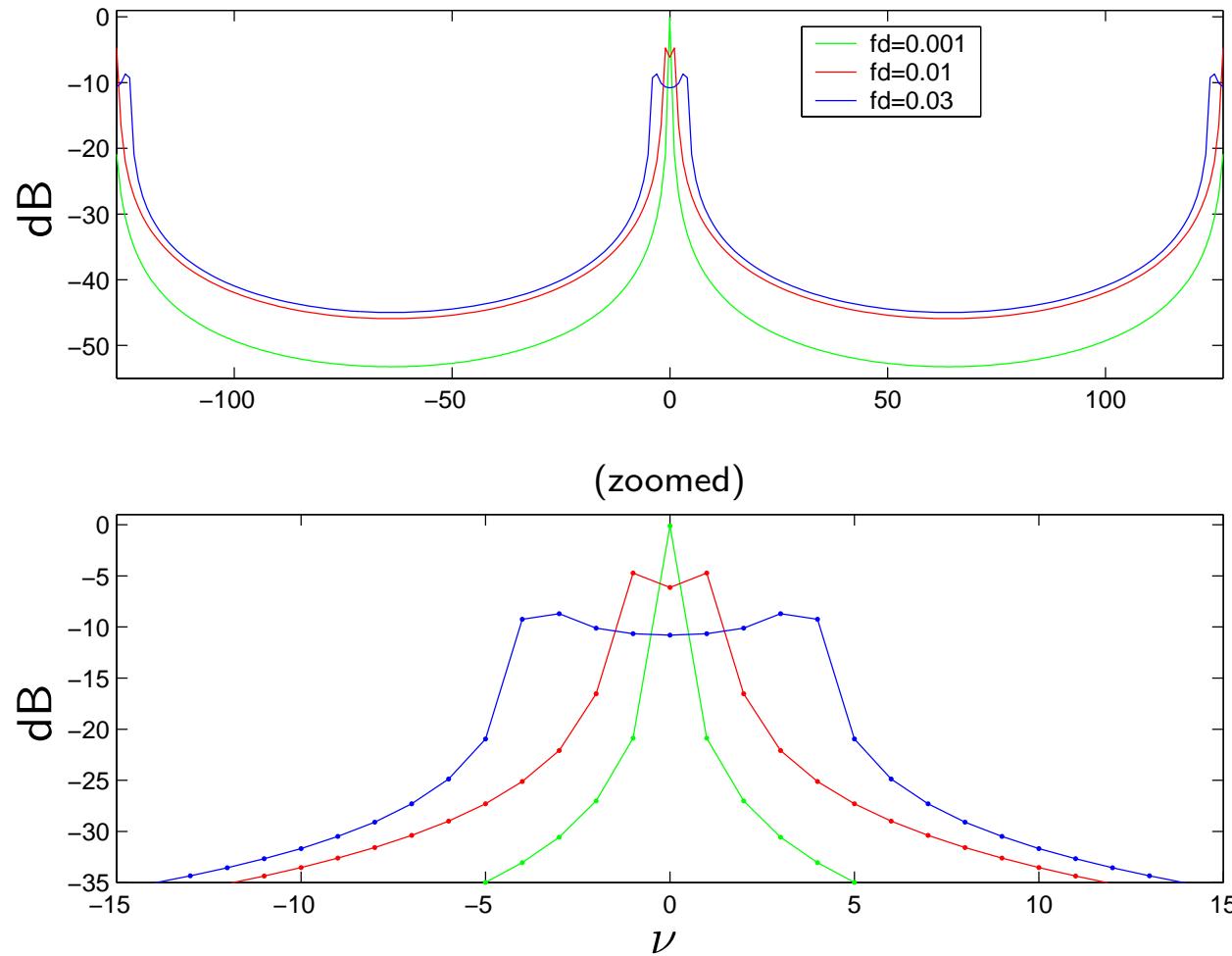
ICI: Doppler Spread meets Finite Block Length:

$$\mathbb{E}\{|h_{\text{df}}(\nu, k)|^2\} = \underbrace{\left(\frac{I_{[0, 2\pi f_d)}(|\phi|) \sum_l \sigma_l^2}{\sqrt{(2\pi f_d)^2 - \phi^2}} \right)}_{\text{assuming WSSUS Rayleigh}} * \underbrace{\left(\frac{\sin(\phi N/2)}{N \sin(\phi/2)} \right)^2}_{\text{block length } N} \Big|_{\phi = \frac{2\pi}{N} \nu}$$



Note: *Zero Doppler spread \Rightarrow Sample at sinc nulls \Rightarrow Zero ICI.*

Rayleigh $E\{|h_{df}(\nu, \cdot)|^2\}$ for $N = 128$ and various f_d :



CP-OFDM Equalization/Detection:

Objective: Recover finite-alphabet vector s from $\boxed{\boldsymbol{x} = \mathcal{H}_{\text{df}} \boldsymbol{s} + \boldsymbol{w}}$.

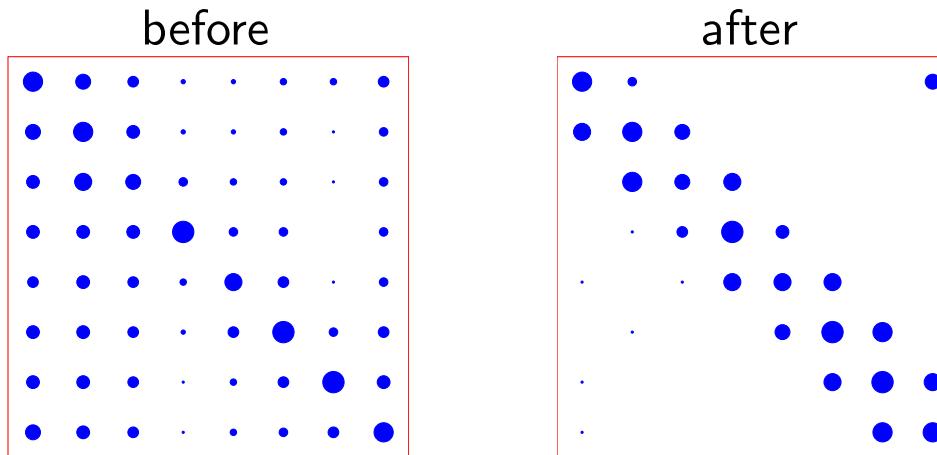
Classical Strategies:

- ZF, LS: $\hat{\boldsymbol{s}}_{\text{zf}} = \text{slice}\left[\mathcal{H}_{\text{df}}^{-1} \boldsymbol{x}\right]$
- MMSE: $\hat{\boldsymbol{s}}_{\text{mmse}} = \text{slice}\left[\mathcal{H}_{\text{df}}^H (\mathcal{H}_{\text{df}} \mathcal{H}_{\text{df}}^H + \sigma_w^2 \boldsymbol{I})^{-1} \boldsymbol{x}\right]$
- MLSD: $\hat{\boldsymbol{s}}_{\text{mlsd}} = \arg \max_{\boldsymbol{s}} \|\boldsymbol{x} - \mathcal{H}_{\text{df}} \boldsymbol{s}\|^2$

LTV channel: \rightsquigarrow Equalization requires $\geq O(N^3)$ operations
 \rightsquigarrow Low-complexity advantage of CP-OFDM is lost!

Linear Pre-Processing to Simplify Detection:

- Use linear pre-processing to simplify detection.
 - Want to make \mathcal{H}_{df} sparse
 - ICI-response “shortening”
 - Reminiscent of ISI-shortening for single-carrier MLSD
- Time-domain windowing = Doppler-domain convolution!



Low-Complexity Pre-Processing = Windowing:

- Apply time-domain window \mathbf{b} before receiver's FFT:

$$\check{\mathbf{x}} = \mathbf{F} \mathcal{D}(\mathbf{b}) \mathbf{r}$$

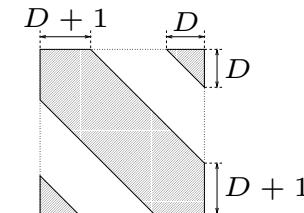
- Equivalent to (circularly) filtering the columns of \mathcal{H}_{df} . . .

$$\begin{aligned}\check{\mathbf{x}} &= \underbrace{\mathbf{F} \mathcal{D}(\mathbf{b}) \mathbf{F}^H}_{\mathcal{C}(\boldsymbol{\beta})} \underbrace{\mathbf{F} \mathbf{r}}_{\mathbf{x}} \\ &= \mathcal{C}(\boldsymbol{\beta}) (\mathcal{H}_{\text{df}} \mathbf{s} + \mathbf{w}) \quad \text{where } \boldsymbol{\beta} = \frac{1}{\sqrt{N}} \mathbf{F} \mathbf{b} \\ &= \underbrace{\mathcal{C}(\boldsymbol{\beta}) \mathcal{H}_{\text{df}}}_{\text{effective channel}} \mathbf{s} + \underbrace{\mathcal{C}(\boldsymbol{\beta}) \mathbf{w}}_{\text{colored noise}}\end{aligned}$$

- This $\mathcal{O}(N)$ processing cannot *perfectly* suppress ICI, but it can come close. . .

Max-SINR Window Coefficients:

- Say we allow $2D$ diagonals of controlled ICI.



- Max-SINR window coefficients \mathbf{b}_* are

$$\mathbf{b}_* = \text{gen-evec}_{\max} \left(\mathbf{A} \odot \mathbf{R}^*, \text{diag}(\mathbf{R} + \sigma^2 \mathbf{I}) - \mathbf{A} \odot \mathbf{R}^* \right)$$

where, for WSSUS Rayleigh fading,

$$[\mathbf{A}]_{m,n} = \frac{\sin\left(\frac{\pi}{N}(2D+1)(n-m)\right)}{N \sin\left(\frac{\pi}{N}(n-m)\right)}$$

$$[\mathbf{R}]_{n,m} = J_0\left(2\pi f_d(n-m)\right) \sum_{l=0}^{N_h-1} \sigma_l^2$$

- Note that \mathbf{b}_* is a function of $\left\{ D, N, f_d, \text{SNR} = \frac{\sum \sigma_l^2}{\sigma^2} \right\}$

Symbol Estimation:

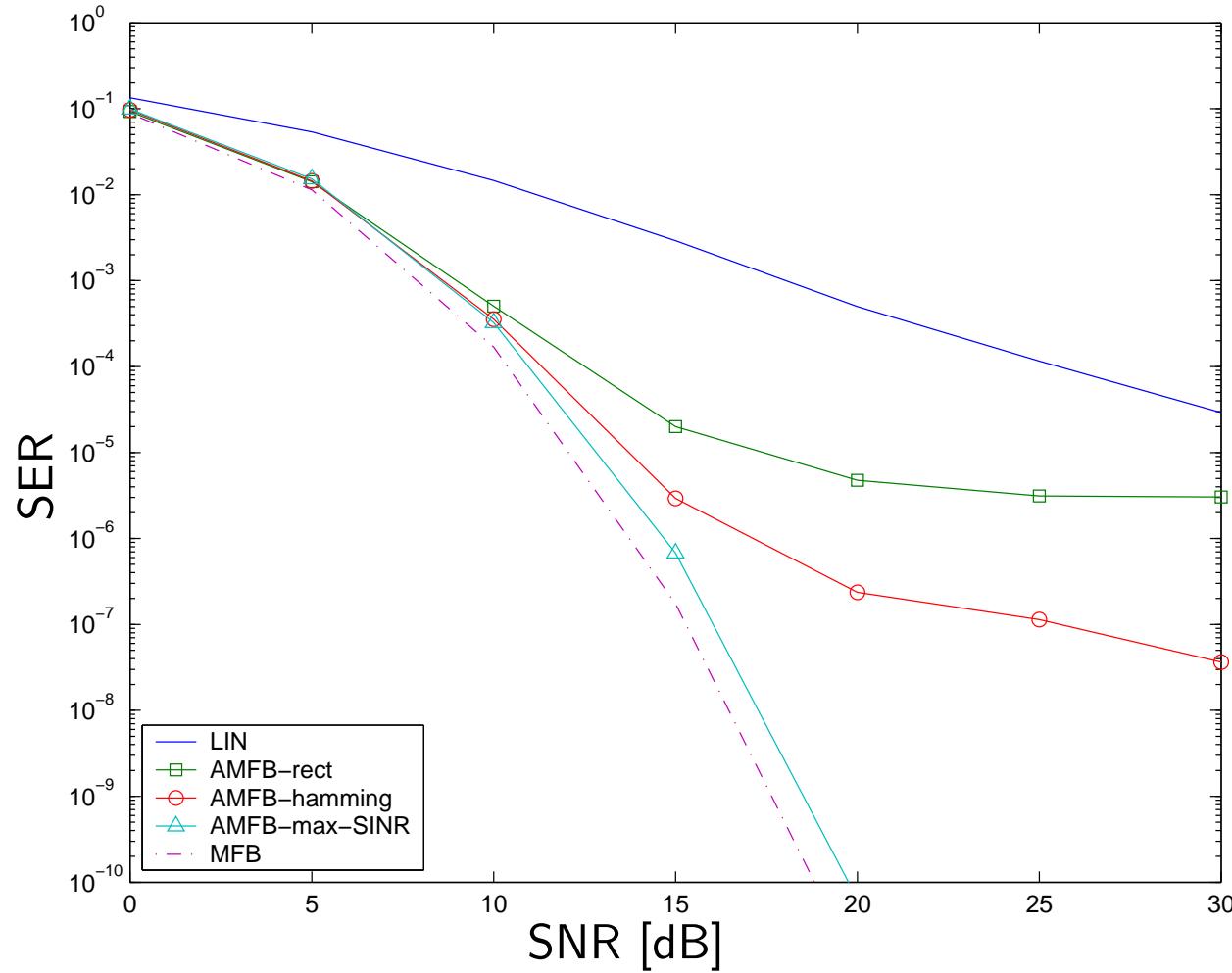
- After windowing to preserve $2D + 1$ diagonals, have

$$\breve{x} = \underbrace{\mathcal{M}_D(\mathcal{C}(\beta)\mathcal{H}_{df}) s}_{\mathcal{H}_{df}} + \underbrace{\overline{\mathcal{M}}_D(\mathcal{C}(\beta)\mathcal{H}_{df}) s}_{\approx 0} + \mathcal{C}(\beta)w$$

where $\mathcal{M}_D(\cdot)$ is a mask operator and $\overline{\mathcal{M}}_D(\cdot)$ its complement.

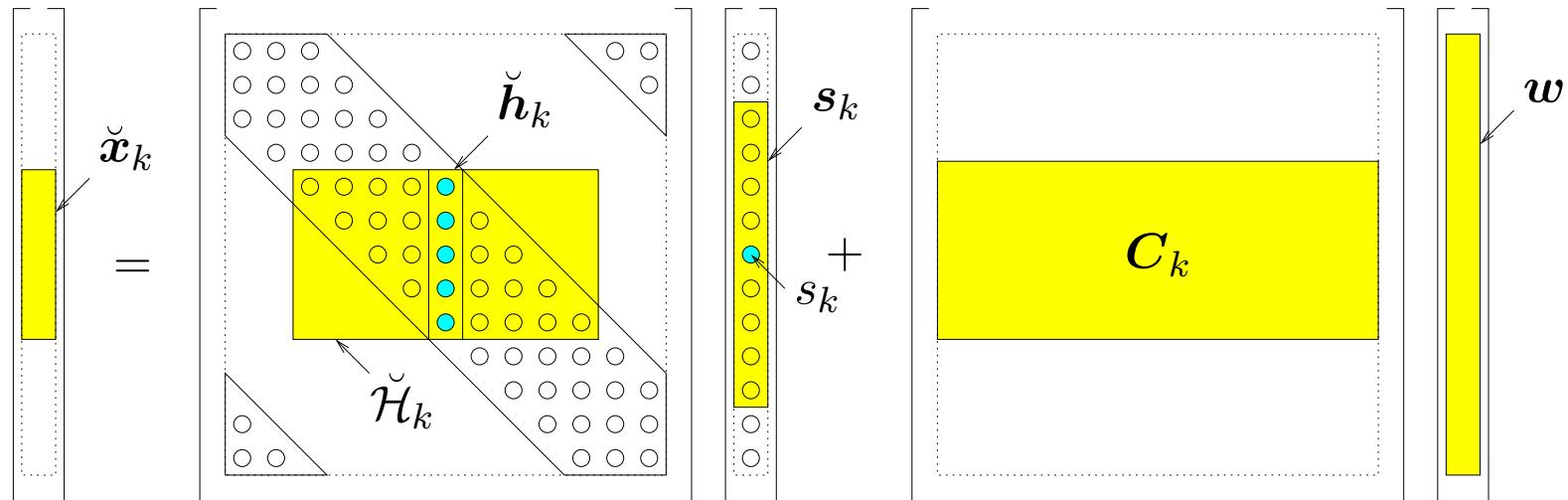
- Goal: Estimate $\{s_0, \dots, s_{N-1}\}$ given \mathcal{H}_{df} , $\mathcal{C}(\beta)$, and \breve{x} .
- Benchmarks:
 - Linear MMSE.
 - MFB: uses true $\mathcal{C}(\beta)\mathcal{H}_{df}$ with perfect interference cancellation.
 - AMFB: uses sparse $\breve{\mathcal{H}}_{df}$ with perfect interference cancellation.

Effect of Windowing on AMFB

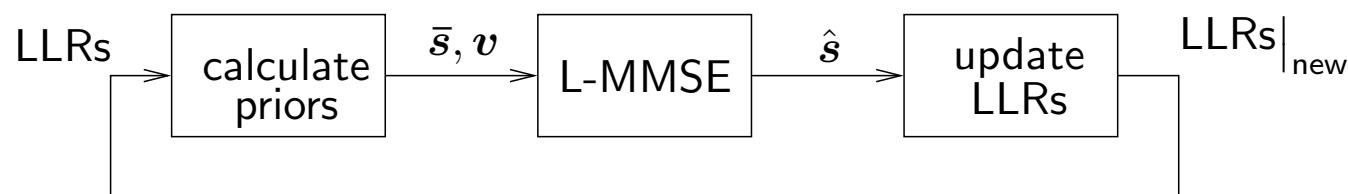


Iterative MMSE Estimation:

L-MMSE estimate of s_k :



Block Iteration for \hat{s} :



Variations on Iterative Joint Estimation:

1. Block Iterative Estimation (BIE):

Estimate block of symbols $\{\hat{s}_0, \dots, \hat{s}_{N-1}\}$ then update block of priors $\{\bar{s}_0, \dots, \bar{s}_{N-1}\}$ and $\{v_0, \dots, v_{N-1}\}$. Repeat...

2. Sequential Iterative Estimation (SIE):

For each k , compute \hat{s}_k then immediately update priors \bar{s}_k and v_k .
Repeat with $k \rightarrow \langle k + 1 \rangle_N \dots$

3. Block Decision Feedback (BDF):

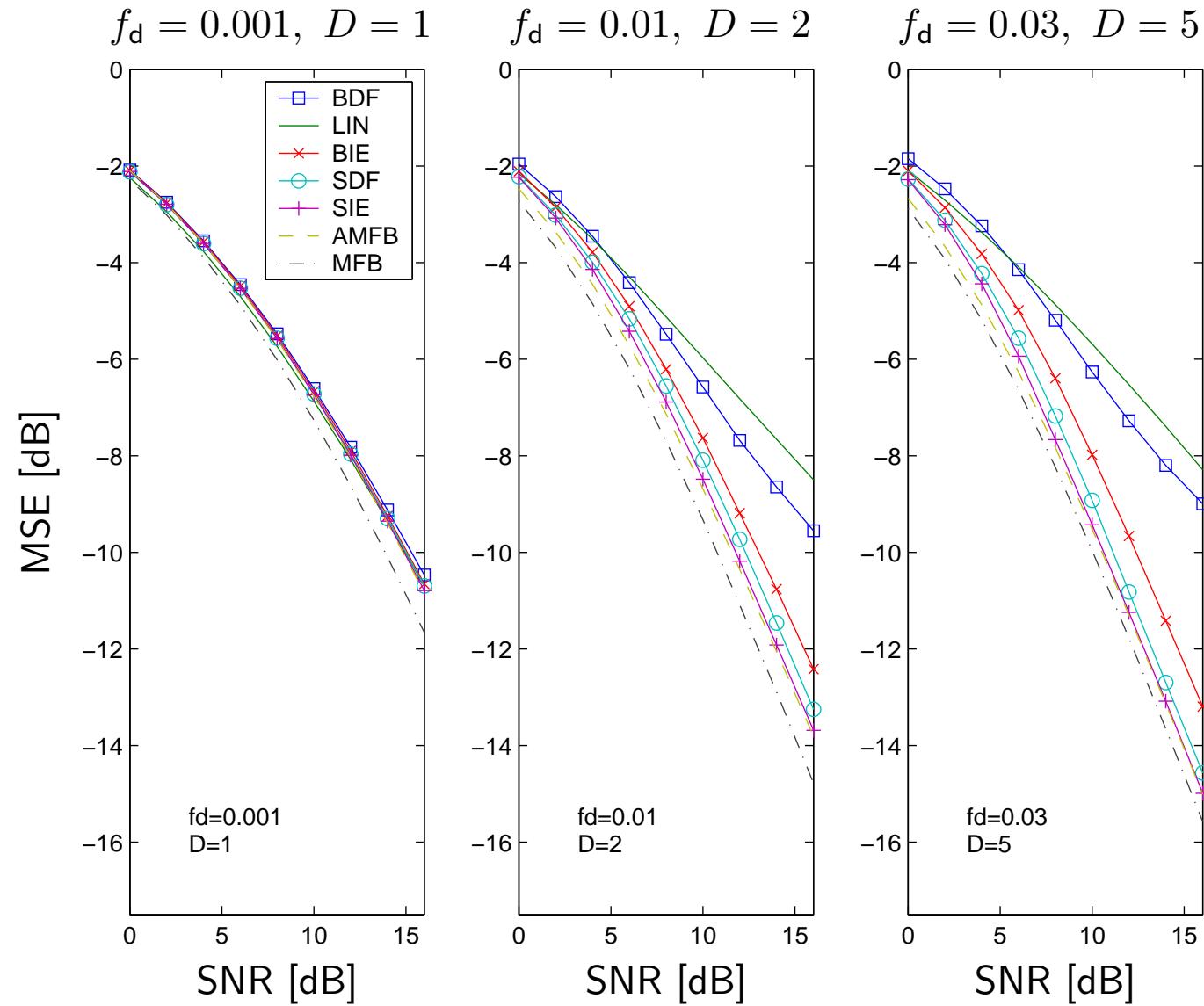
Like BIE, but $\bar{s}_k^{(i+1)} = \text{sgn}(\hat{s}_k^{(i)})$ and $v_k^{(i+1)} = 0$.

4. Sequential Decision Feedback (SDF):

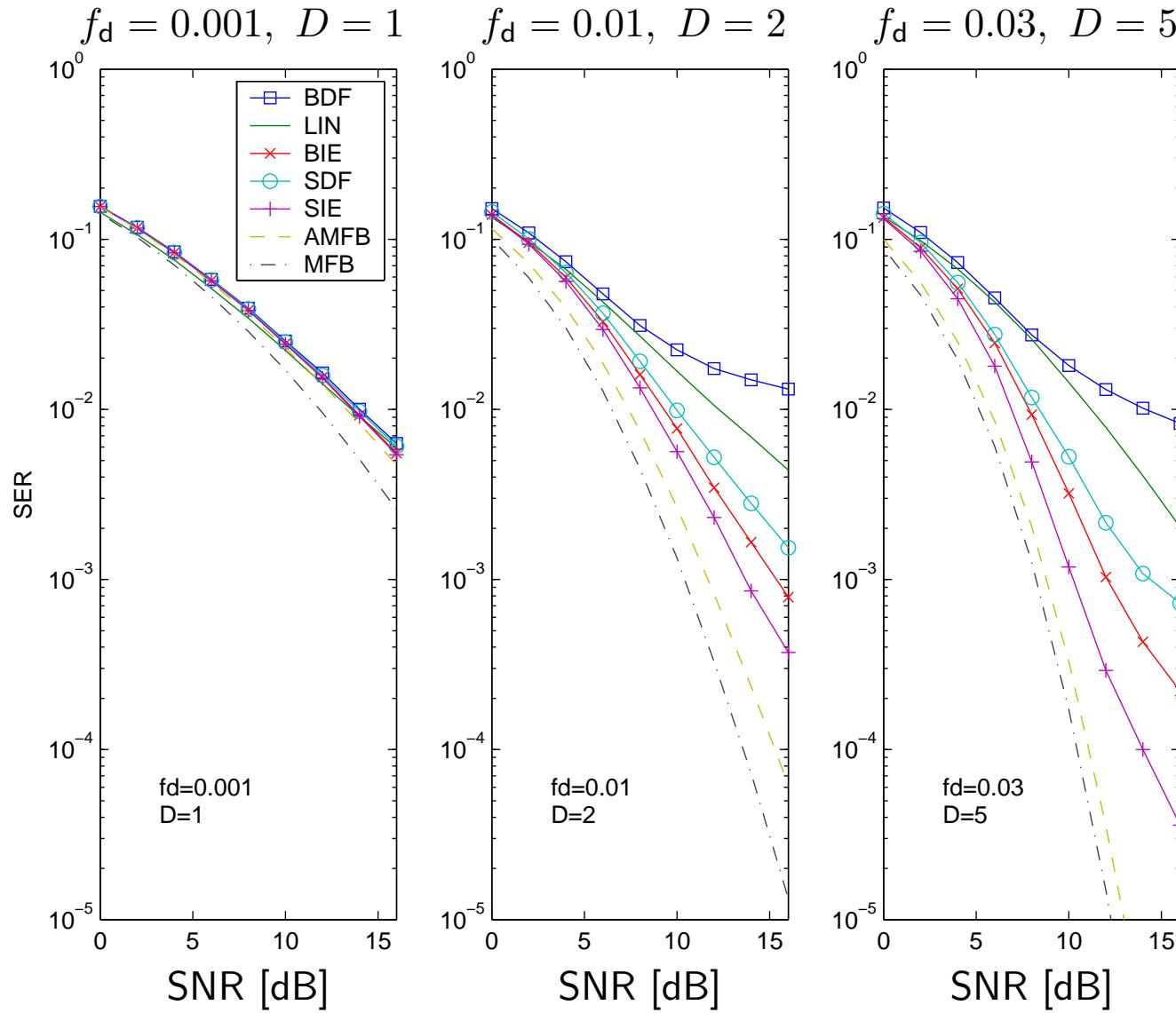
Like SIE, but $\bar{s}_k^{(i+1)} = \text{sgn}(\hat{s}_k^{(i)})$ and $v_k^{(i+1)} = 0$.

Exploit banded $\check{\mathcal{H}}_{\text{df}}$; Need only $\mathcal{O}(D^2 N)$ operations per iteration.

MSE versus SNR after 2 iterations ($N = 128, b = \hat{b}_\star$) :



SER versus SNR after 2 iterations ($N = 128, b = \hat{b}_\star$) :



Summary:

- CP-OFDM reception complicated by time-selectivity.
- Proposed a two-stage CP-OFDM receiver for doubly-selective channels:
 1. SINR-optimal windowing,
 2. Iterative MMSE estimation.
- Like classical CP-OFDM receivers, requires $\mathcal{O}(\log N)$ ops/symbol.
- Performance:
 - MSE is $\sim 1\text{--}2\text{dB}$ from MFB.
 - Uncoded error rate is $\sim 3\text{dB}$ from MFB.
 - Soft decoding can be easily incorporated in increase performance.