



A Factor-Graph Approach to Joint OFDM Channel Estimation and Decoding in Impulsive Noise Channels

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Outline

- Uncoordinated interference in communication systems
- Effect of interference on OFDM systems
- Prior work on OFDM receivers in uncoordinated interference
- Message-passing OFDM receiver design
- Simulation results

Uncoordinated Interference

- Typical Scenarios:
 - Wireless Networks:
 Ad-hoc Networks, Platform Noise, non-communication sources
 - Powerline Communication Networks:
 Non-interoperable standards, electromagnetic emissions
- Statistical Model:

Gaussian Mixture (GM)

$$p(n) = \sum_{k=0}^{K-1} \pi_k \mathcal{N}(n; 0, \gamma_k)$$

 $K \in \mathbb{N}$:# of comp.

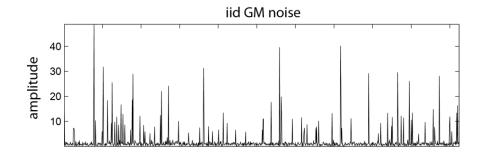
 π_i : comp. probability

 γ_i : comp. variance

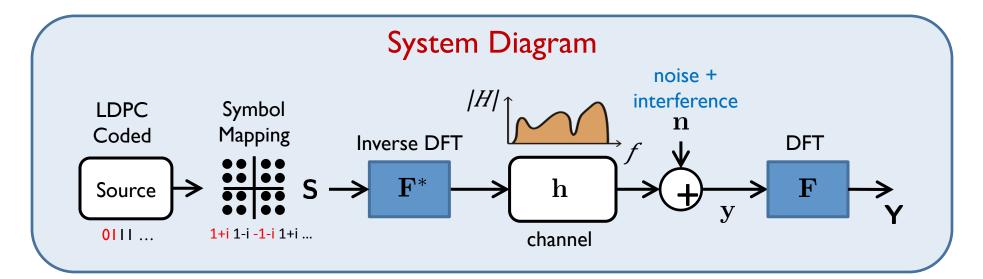
Interference Model

Two impulsive components:

- 7% of time/20dB above background
- 3% of time/30dB above background



OFDM Basics



Noise Model

$$n_j = g_j + i_j$$

 n_i : total noise

where g_j : background noise

 i_j : interference

and $g_j \sim \mathcal{N}(0, N_0)$

 i_j : GM or GHMM

Receiver Model

After discarding the cyclic prefix:

$$y = Hs + n = HF^*S + g + i$$

After applying DFT:

$$Y = Fy = FHF^*S + Fn = H \circ S + G + I$$

• Subchannels: $Y_k = H_k S_k + G_k + I_k$

OFDM Symbol Structure

Coding

 Added redundancy protects against errors



Data Tones

$$\mathsf{Y}_k = \mathsf{H}_k \mathsf{S}_k + \mathsf{G}_k + \mathsf{I}_k$$

- Symbols carry information
- Finite symbol constellation
- Adapt to channel conditions

Pilot Tones

$$\mathsf{Y}_k = \mathsf{H}_k \mathsf{p} + \mathsf{G}_k + \mathsf{I}_k$$

- Known symbol (p)
- Used to estimate channel

Null Tones

$$\mathsf{Y}_k = \mathsf{G}_k + \mathsf{I}_k$$

- Edge tones (spectral masking)
- Guard and low SNR tones
- Ignored in decoding

pilots \rightarrow linear channel estimation \rightarrow symbol detection \rightarrow decoding

But, there is unexploited information and dependencies

Prior OFDM Designs

Category	References	Method	Limitations
Time-Domain preprocessing (PP)	[Haring2001]	Time-domain signal estimation	 ignore OFDM signal structure performance degrades with increasing SNR and modulation order
	[Zhidkov2008, Tseng2012]	Time-domain signal thresholding	
Sparse Signal Reconstruction	[Caire2008, Lampe2011]	Compressed sensing	utilize only known tonesdon't use interference modelscomplexity
	[Lin2011]	Sparse Bayesian Learning (SBL)	
Iterative Receivers	[Mengi2010, Yih2012]	Iterative preprocessing & decoding	suffer from preprocessing limitationsad-hoc design
	[Haring2004]	Turbo-like receiver	

All don't consider the non-linear channel estimation, and don't use code structure

Joint MAP-Decoding

The MAP decoding rule of LDPC coded OFDM is:

$$\underset{b_m \in \{0,1\}}{\operatorname{arg\,max}} P(b_m | \mathbf{Y}; \Theta) \quad \forall m$$

Can be computed as follows:

$$P(b_m|\mathbf{Y};\Theta) = \sum_{\mathbf{b}_{\backslash \mathbf{m}}} P(\mathbf{b}|\mathbf{Y};\Theta) \propto \sum_{\mathbf{b}_{\backslash \mathbf{m}}} p(\mathbf{Y}|\mathbf{b};\Theta) P(\mathbf{b})$$

$$\propto \sum_{\mathbf{S},\mathbf{c},\mathbf{b}_{\backslash \mathbf{m}}} \prod_{k=0}^{N-1} \underbrace{\int_{\mathbf{i},\mathbf{h}} p(\mathbf{Y}_k|\mathbf{S}_k,\mathbf{h},\mathbf{i};\Theta) p(\mathbf{h};\theta_{\mathbf{h}}) p(\mathbf{i};\theta_{\mathbf{i}}) P(\mathbf{S}|\mathbf{c}) P(\mathbf{c}|\mathbf{b})}_{\text{depends on linearly-mixed N noise non iid & LDPC code samples and L channel taps non-Gaussian}}$$

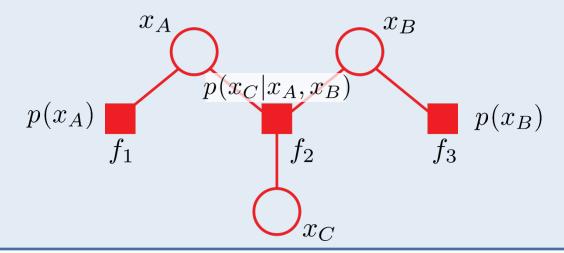
Very high dimensional integrals and summations !!

Belief Propagation on Factor Graphs

- Graphical representation of pdf-factorization
- Two types of nodes:
 - variable nodes denoted by circles
 - factor nodes (squares): represent variable "dependence"
- Consider the following pdf:

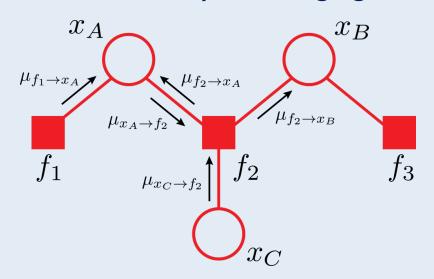
$$p(x_A, x_B, x_C) = p(x_A)p(x_B)p(x_C|x_A, x_B)$$

• Corresponding factor graph:



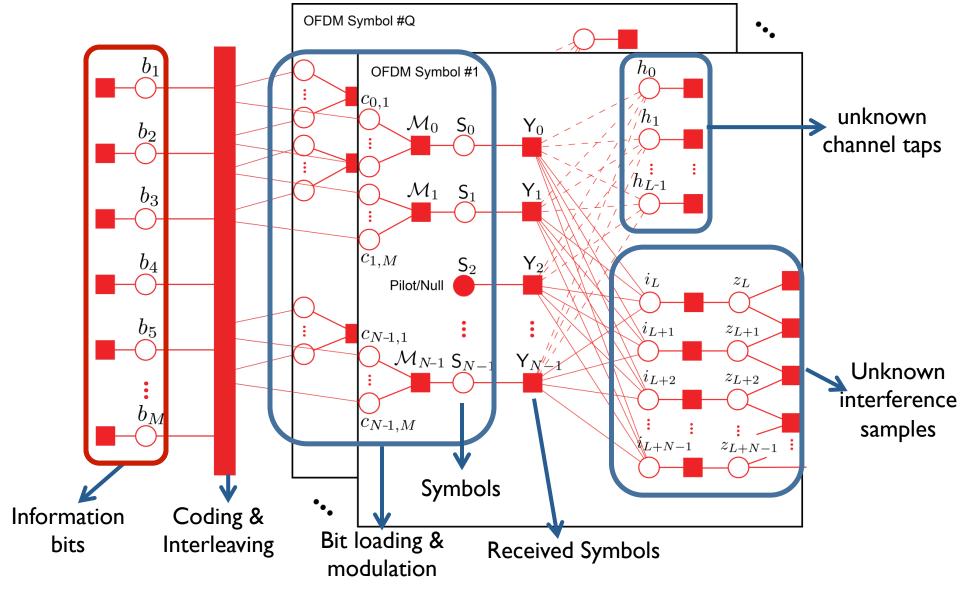
Belief Propagation on Factor Graphs

Approximates MAP inference by exchanging messages on graph

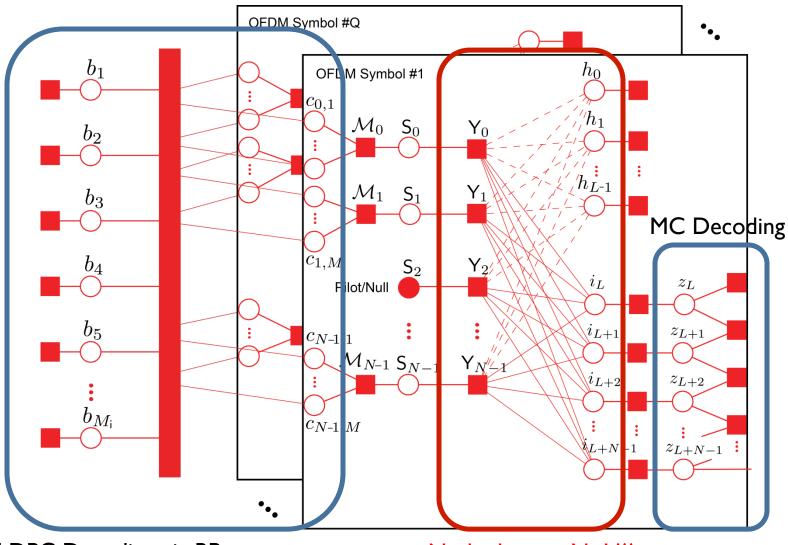


- Factor message = factor's belief about a variable's p.d.f.
- Variable message = variable's belief about its own p.d.f.
- Variable operation = multiply messages to update p.d.f.
- Factor operation = merges beliefs about variable and forwards
- Complexity = number of messages = node degrees

Coded OFDM Factor Graph



BP over **OFDM** Factor Graph



LDPC Decoding via BP [MacKay2003]

Node degree=N+L!!!

Generalized Approximate Message Passing

[Donoho2009,Rangan2010]

Estimation with Linear Mixing

observations

variables

$$y_1 \overset{p(y_1|z_1)}{\longleftarrow} z_1 \overset{\text{coupling}}{\longleftarrow} x_1 \overset{p(x_1)}{\longleftarrow} x_1 \overset{p(x_1)$$

- Generally a hard problem due to coupling
- Regression, compressed sensing, ...
- **OFDM** systems:

Interference subgraph

channel subgraph

given **H**

given

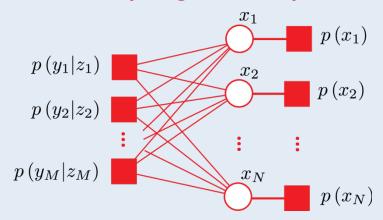
 $\mathbf{z} = \mathbf{I}$ and $\mathbf{x} = \mathbf{i}$ $\mathbf{z} = \mathbf{H}$ and $\mathbf{x} = \mathbf{h}$

 $\Phi = \mathbf{F}$

 $\mathbf{\Phi} = \sqrt{N} \mathbf{F}_{1:L}$

3 types of output channels for each

Decoupling via Graphs

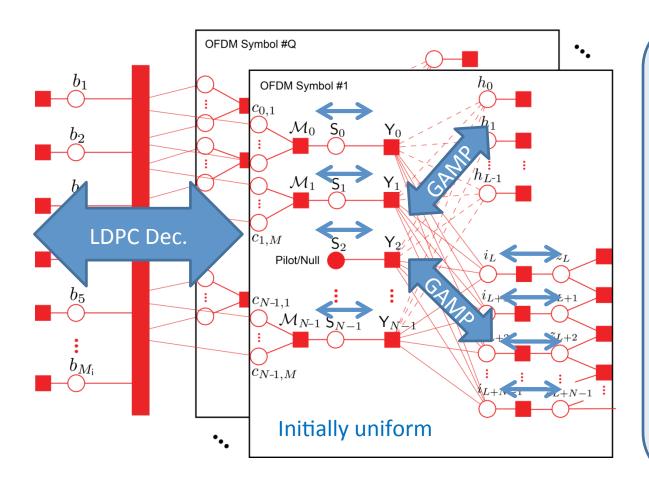


- If graph is sparse use standard BP
- If dense and "large" →

Central Limit Theorem

- At factors nodes treat z as Normal
- Depend only on means and variances of incoming messages
- Non-Gaussian output \rightarrow quad approx.
- Similarly for variable nodes
- Series of scalar MMSE estimation problems: O(N+M) messages

Message-Passing Receiver



Schedule

Turbo Iteration:

- I. coded bits to symbols
- 2. symbols to \mathbf{Y}
- 3. Run channel GAMP
- 4. Run noise "equalizer"
- 5. **Y** to symbols
- 6. Symbols to coded bits
- 7. Run LDPC decoding

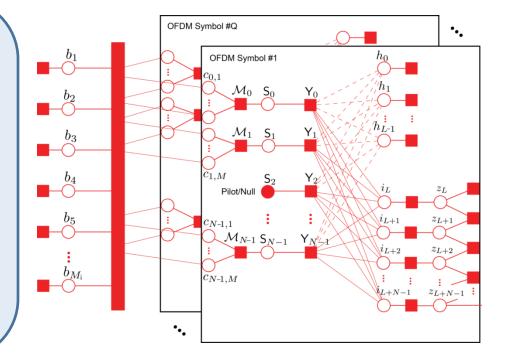
Equalizer Iteration:

- I. Run noise GAMP
- 2. MC Decoding
- 3. Repeat

Receiver Design & Complexity

Design Freedom

- Not all samples required for sparse interference estimation
- Receiver can pick the subchannels:
 - Information provided
 - Complexity of MMSE estimation
- Selectively run subgraphs
 - Monitor convergence (GAMP variances)
 - Complexity and resources
- GAMP can be parallelized effectively



Operation	Complexity per iteration	
MC Decoding	$\mathcal{O}(N)$	
LDPC Decoding	$\mathcal{O}(M_c+C)$	
GAMP	$\mathcal{O}(\min[N\log N, U ^2])$	

Notation

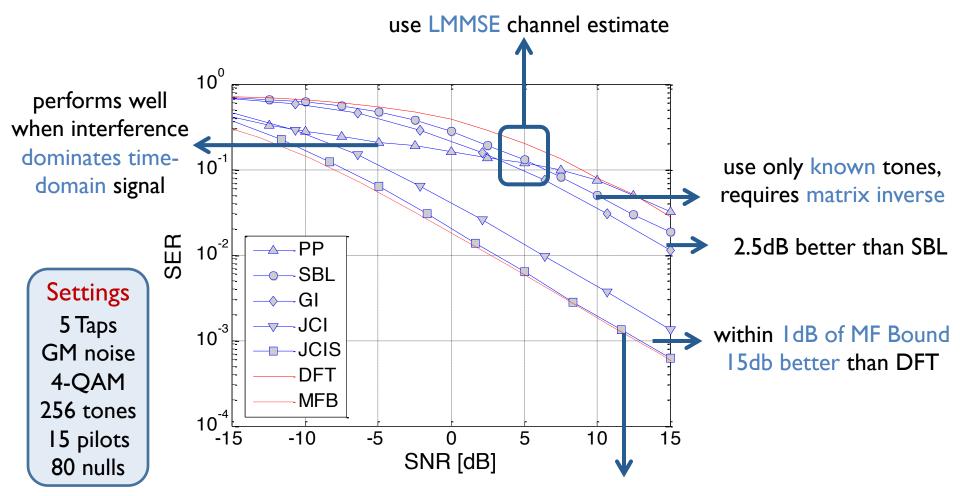
 $N: \# \ {\sf tones}$

 M_c : # coded bits

C:# check nodes

U: set of used tones

Simulation - Uncoded Performance



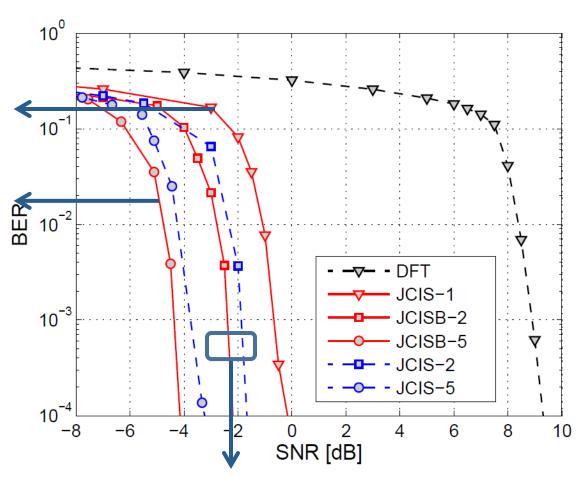
Matched Filter Bound: Send only one symbol at tone k

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Simulation - Coded Performance

one turbo iteration gives 9db over DFT

5 turbo iterations gives 13dB over DFT



Settings

I0 Taps
GM noise
I6-QAM
N=1024
I50 pilots
Rate ½
L=60k

Integrating LDPC-BP into JCNED by passing back bit LLRs gives I dB improvement

Summary

- Huge performance gains if receiver account for uncoordinated interference
- The proposed solution combines all available information to perform approximate-MAP inference
- Asymptotic complexity similar to conventional OFDM receiver
- Can be parallelized
- Highly flexible framework: performance vs. complexity tradeoff