



Signal Detection Based on Belief Propagation in a Massive MIMO System

Takeo Ohgane
Hokkaido University, Japan

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Background (1)

- Massive MIMO

- An order of 100 antenna elements

- channel capacity issue

- Fading reduction effect

H : N x N channel matrix

(i.i.d. complex Gaussian random variables)



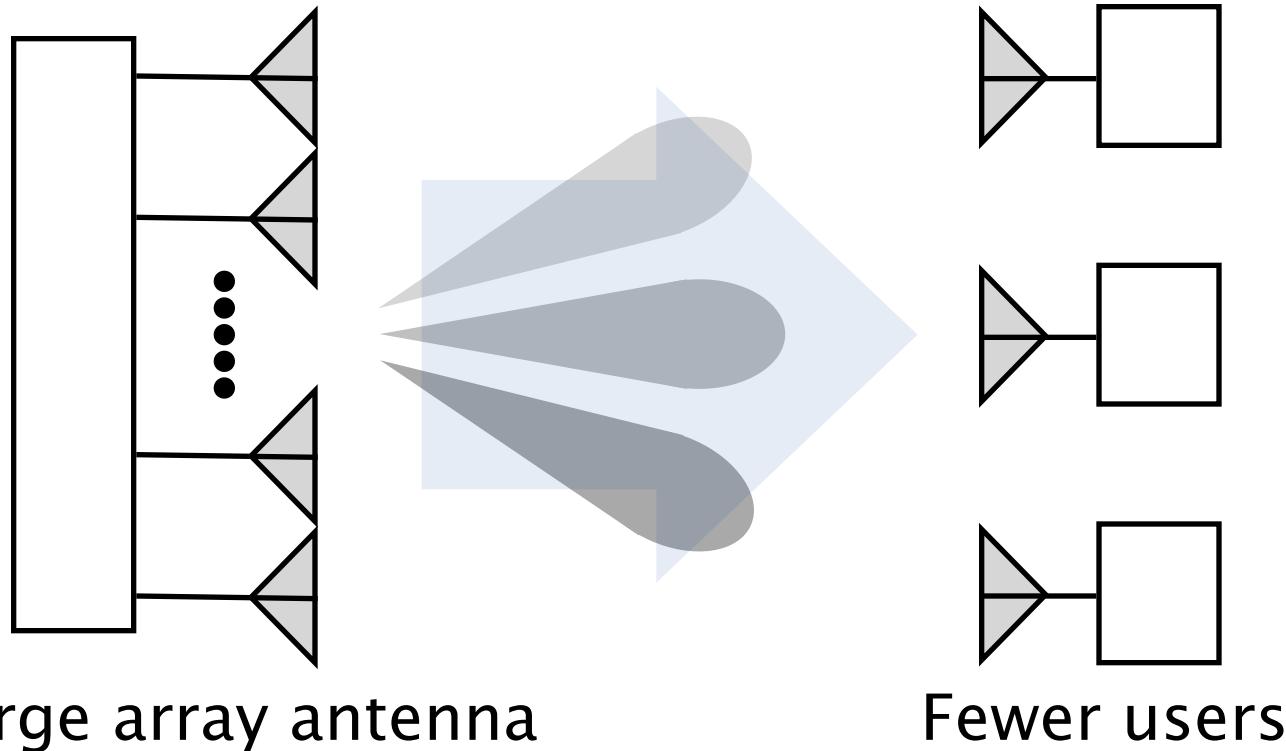
$$H^H H \simeq NI$$

$$\begin{aligned} \log \det \left(I + \frac{\gamma}{N} H^H H \right) &\simeq \log \det (I + \gamma I) \\ &= N \log(1 + \gamma) \end{aligned}$$



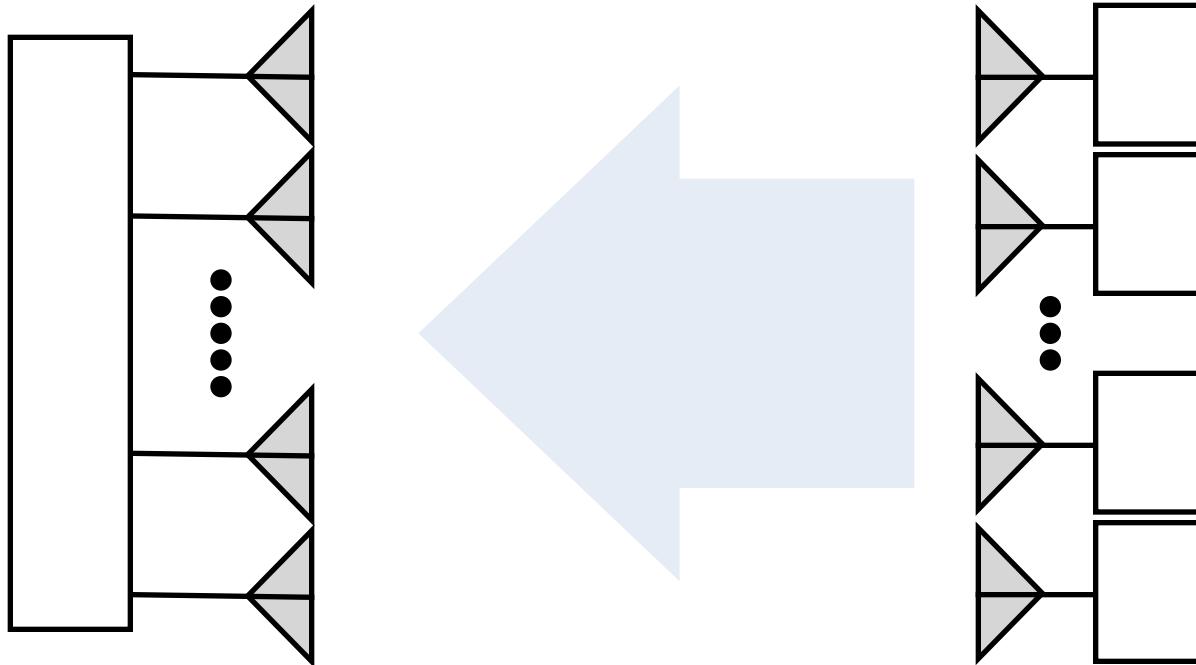
Background (2)

- Promising application
 - Multiuser MIMO



Problem

- If the number of users increase.....



Detection of a large number of streams is required.



Objective

- Spatial demultiplexing

- MAP detection: $O(L^N)$
- Spatial filtering (MMSE, ZF): $O(N^3)$
- QR decomposition based algorithms: $O(N^3)$

- Less complex method

- Detection based on belief propagation (BP): $O(N^2)$

N. Srinidhi, S. K. Mohammed, A. Chockalingam, and B. S. Rajan, “Low-Complexity Near-ML Decoding of Large Non-Orthogonal STBCs using Reactive Tabu Search,” Proc. IEEE ISIT, pp. 1993–1997, June/July 2009.

C. Knievel, M. Noemm, and P. A. Hoeher, “Low-Complexity Receiver for Large-MIMO Space-Time Coded Systems,” Proc. IEEE VTC-Fall, Sept. 2011.

- Capability of pure BP-based algorithm

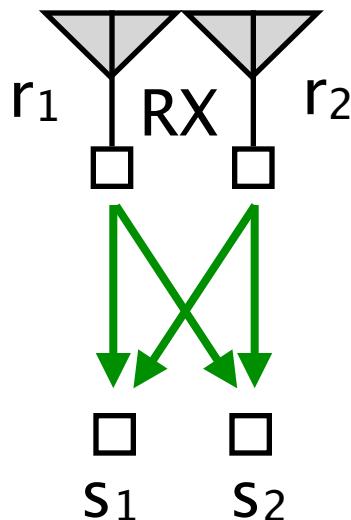
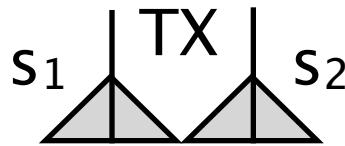


Contents

- Factor graph expression
- BP-based detection algorithm
- EXIT analysis
- Performance evaluation
- Conclusions



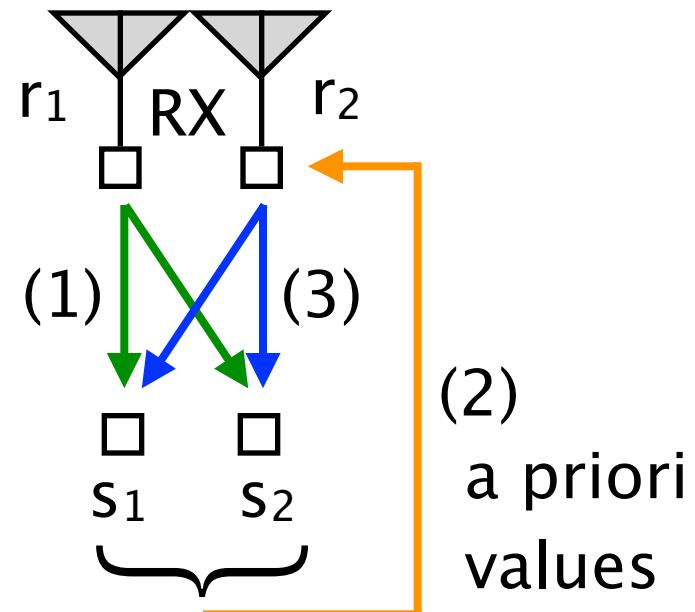
Factor graph expression (1)



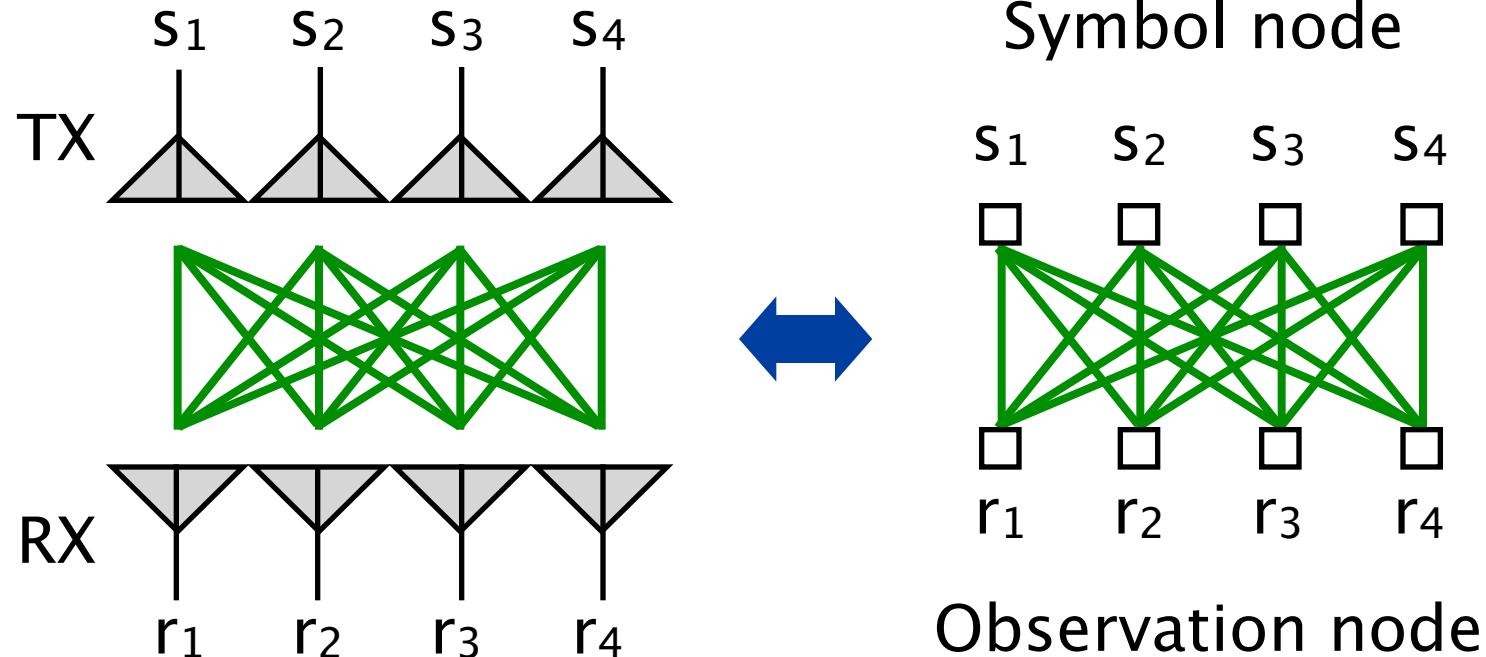
Problem is to estimate s_1 and s_2 from r_1 and r_2 .

Estimated symbols and received signals are mutually related.

ex)

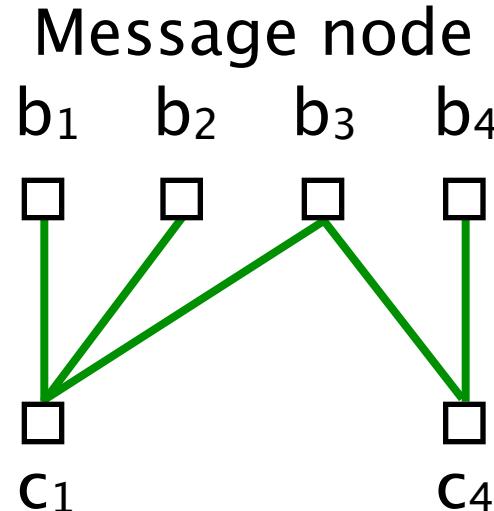


Factor graph expression (2)

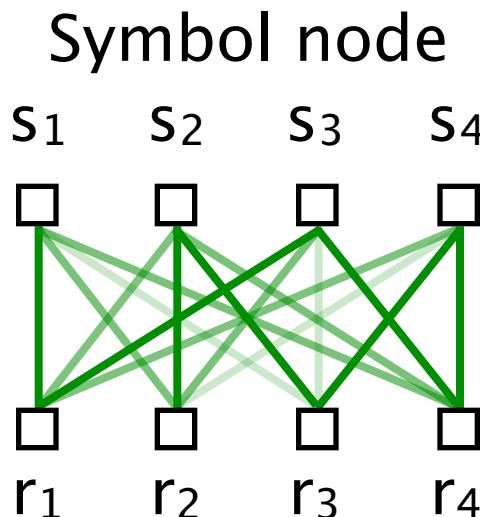


Factor graph expression (3)

- MIMO detection vs LDPC decoding



$$b_1 + b_2 + b_3 = 0 \quad b_3 + b_4 = 0$$



$$r_i = h_{i1}s_1 + h_{i2}s_2 + h_{i3}s_3 + h_{i4}s_4$$



BP-based detection (1)

- Message update at observation node

(1) soft replica generation

$$\beta_{ik} \xrightarrow{\text{blue arrow}} \hat{s}_k$$

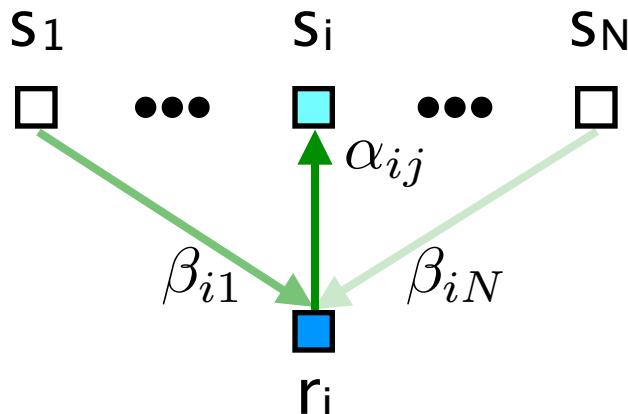
extrinsic value

(2) parallel interference cancellation (PIC)

$$\hat{r}_i^{(j)} = r_i - \sum_{k=1, k \neq j}^N h_{ik} \hat{s}_k \quad O(N^2)$$

(3) detection

$$\log \frac{P(\hat{r}_i^{(j)} | s_j = 1)}{P(\hat{r}_i^{(j)} | s_j = 0)} = \alpha_{ij}$$



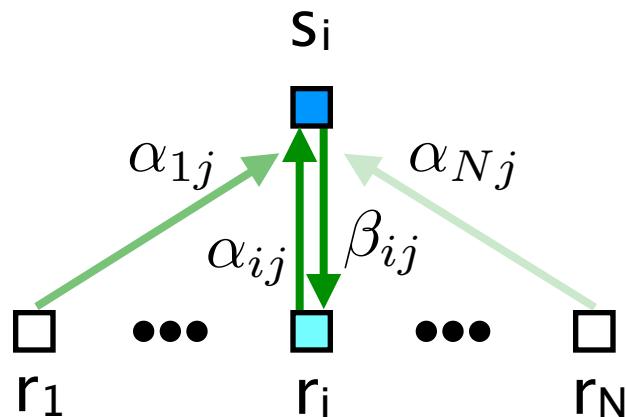
$\alpha_{pq} \ \beta_{pq}$: LLR



BP-based detection (2)

- Message update at symbol node

(1) a posteriori LLR calculation



$$\gamma_j = \sum_{k=1}^N \alpha_{kj}$$

$\gamma_j \rightarrow$ final decision

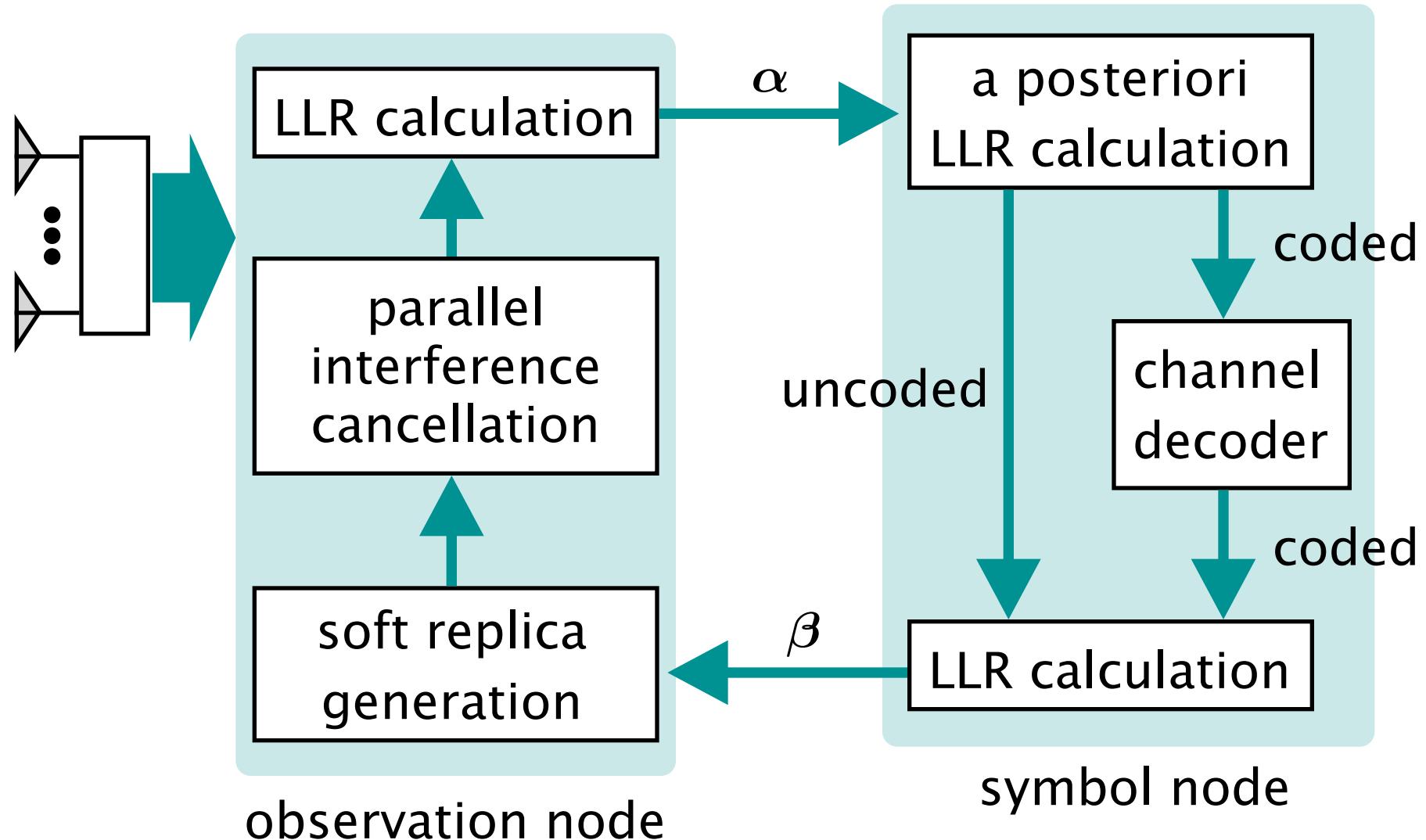
(2) message calculation

$$\underline{\beta_{ij}} = \gamma_j - \alpha_{ij}$$

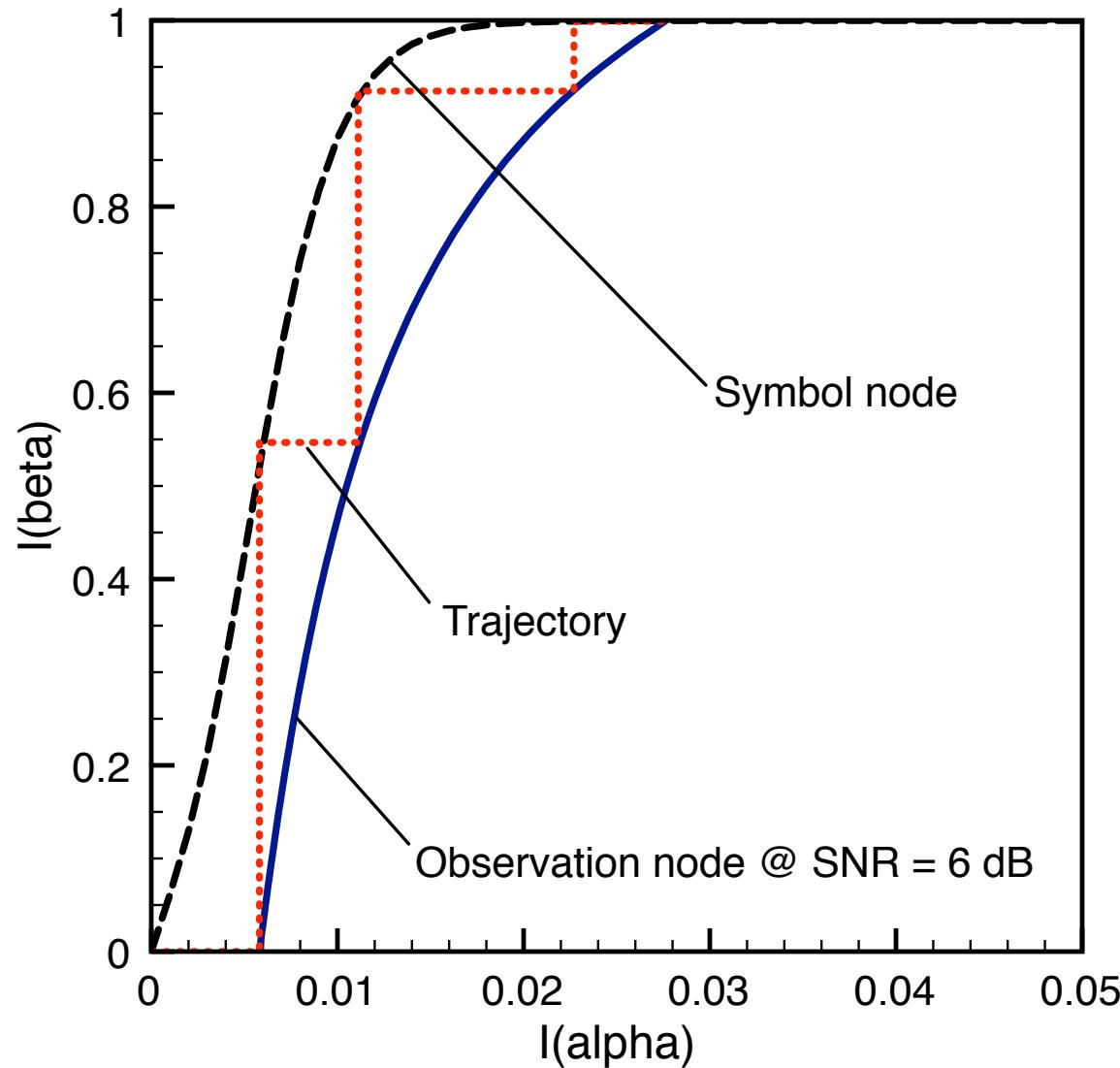
extrinsic value



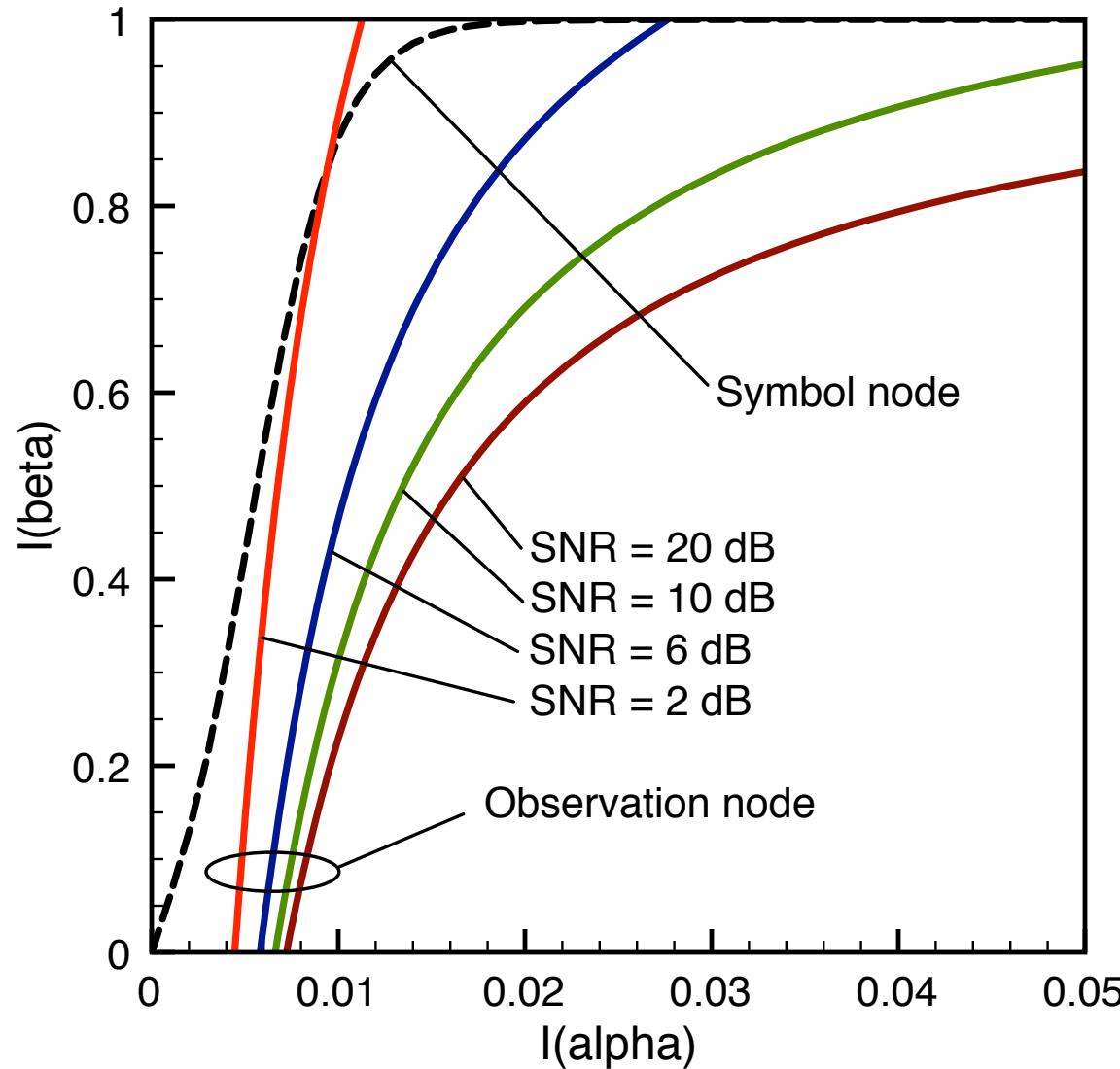
Block diagram of BP-based detection



An example of reliability improvement



EXIT analysis

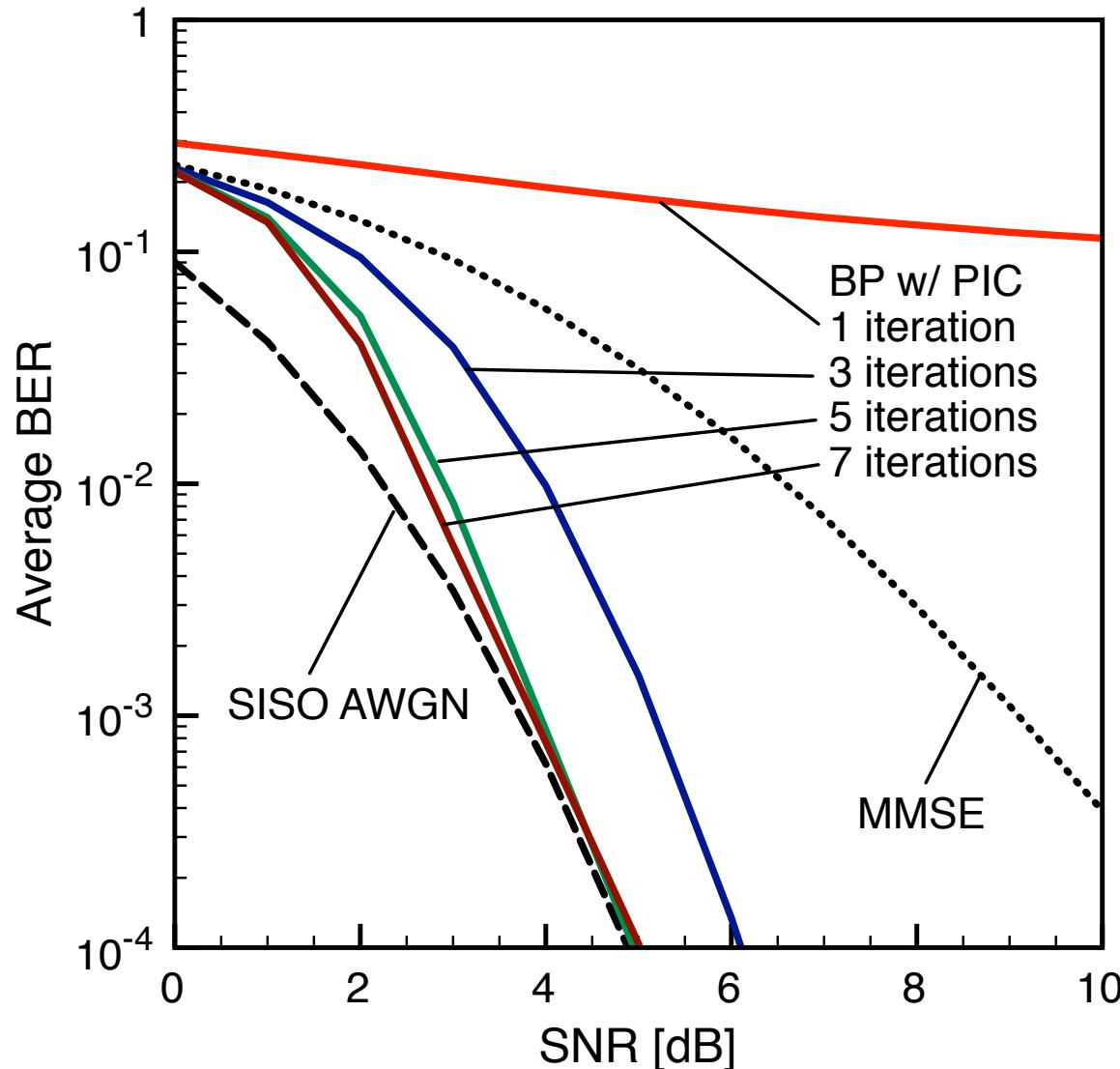


Simulation parameters

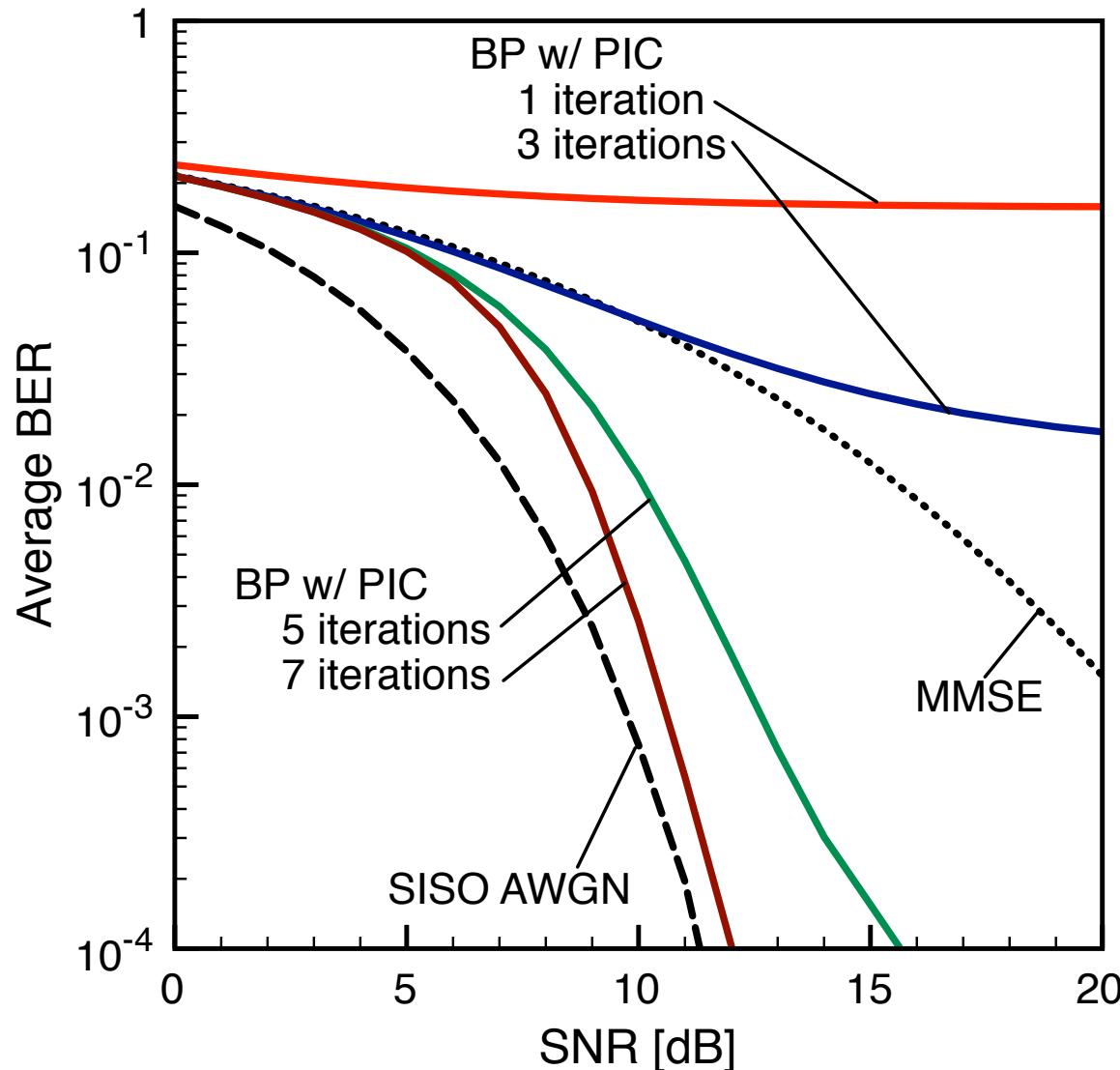
	BP with PIC	MMSE
Number of antennas	$10 \times 10, 30 \times 30, 50 \times 50,$ $100 \times 100, 200 \times 200$	
Modulation		QPSK
Channel statistics		quasi-static Rayleigh fading
Noise		AWGN
Frame length		10–100 symbols
Number of frames		10,000
Channel encoding		Convolutional code (constraint length 3, coding rate 1/2)
Channel decoding		Max-Log MAP decoder



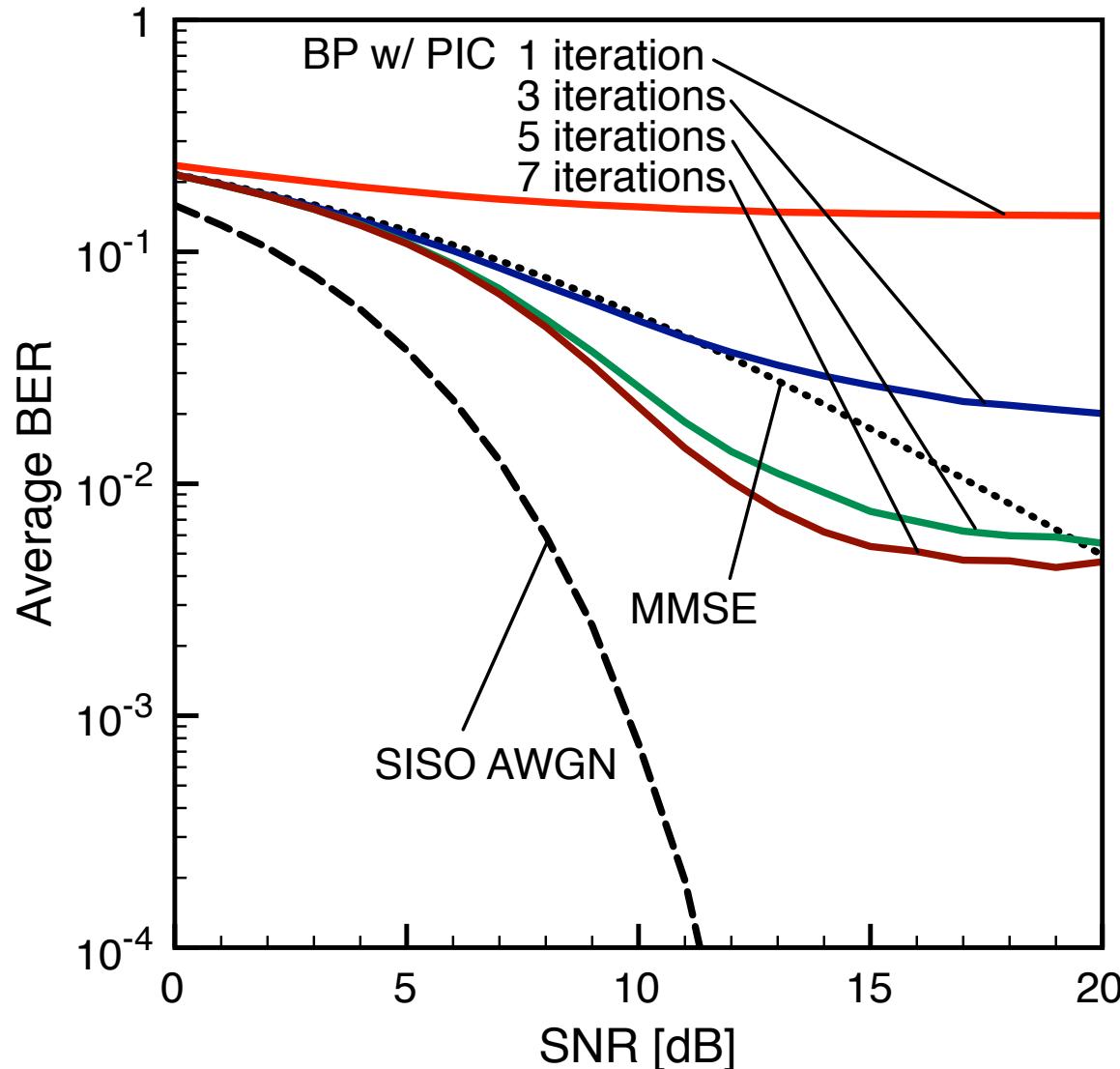
BER performance (100 x 100, coded case)



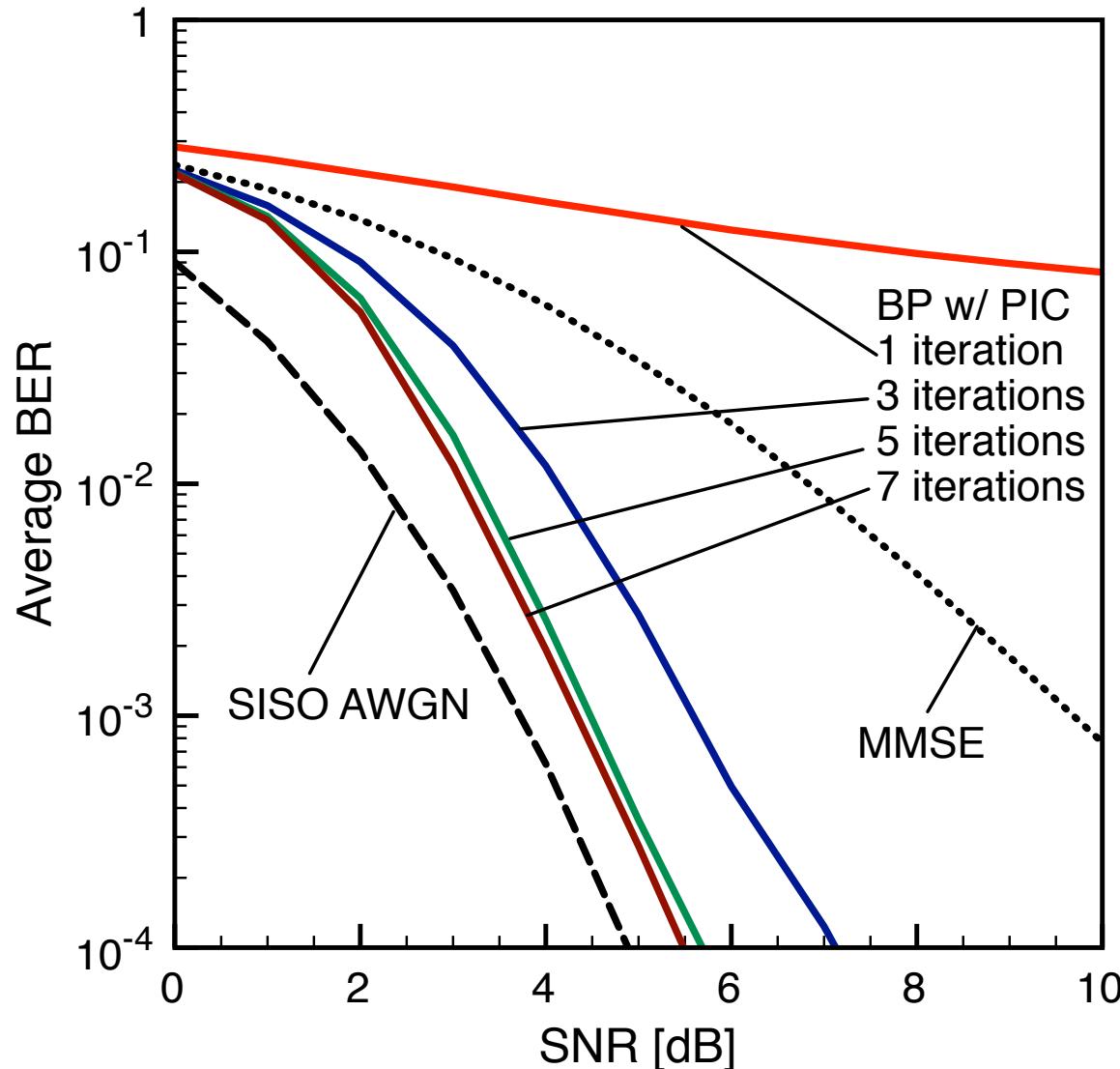
BER performance (100 x 100, uncoded case)



BER performance (10 x 10, uncoded case)



BER performance (10 x 10, coded case)



Conclusions

- We have clarified the capability of pure BP-based detection for spatially multiplexed streams.
- BP-based detection
 - is implementable in $O(N^2)$
 - achieves better performance than spatial filtering with MMSE
 - works only when size of array is large
 - converges more quickly with channel encoding
- Future works
 - effects of channel correlation
 - higher level modulation

