

Sphere Decoding

Application to a MC-CSK system

Sphere Decoding

- Mathematical approach

Minimization of a cost function $f(x_1, \dots, x_K)$ with respect to its K arguments taking value in a discrete set of cardinality L

- Digital communications approach

Minimization of the distance between the received symbol and the possible transmitted symbols in order to recover the information in a multi-user system

Sphere Decoding

- Computation of the distance based on the optimum Maximum Likelihood operator
→ absolute minimum

- Smart hypotheses leading to a smaller complexity than classic ML algorithm

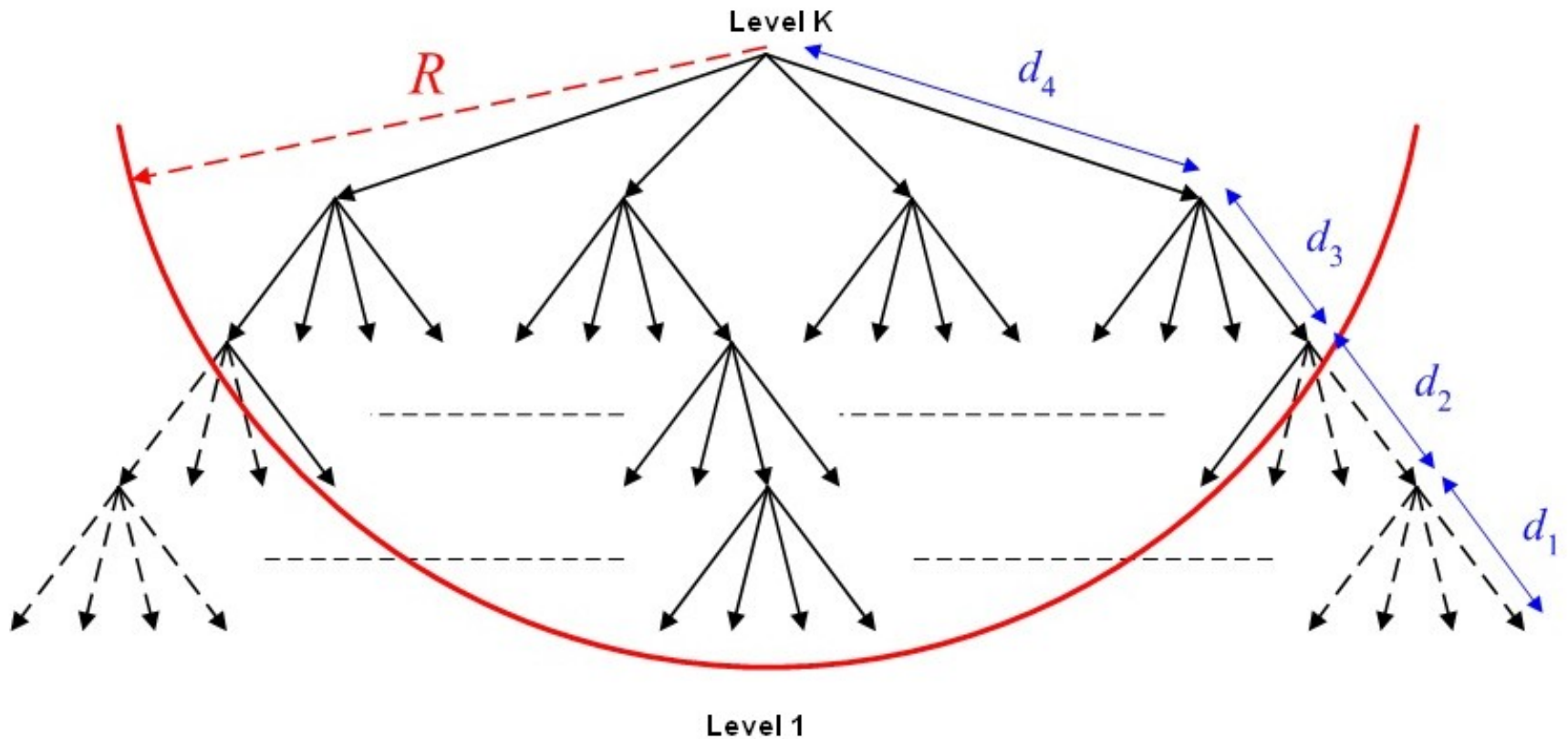
Main idea: searching over the noiseless possible received signals that lie within a hypersphere of radius R around the actual received signal

- Multi-user detection

Principle of Sphere Decoding

- The distance is seen as a sum of nonnegative functions with an increasing number of arguments
$$d = f(x_1, \dots, x_K) = h(x_K) + h(x_K, x_{K-1}) + \dots + h(x_K, \dots, x_1)$$
- Graphically, the process is a (K)-level tree graph with one uppermost node (level K) and L^{K-1} leaves (level 1)
- Each branch corresponds to an intermediate distance

Graphical approach



Algorithm

- 2 stages
 - Initialization
 - Pruning

Algorithm

- 2 stages

- Initialization

- At level $K \rightarrow$ smallest intermediate value leads to L next nodes at level $K-1$.
 - At level $K-1 \rightarrow$ smallest value among the L nodes and so on until the level 1.
 - Sum of the K values gives a first minimum of the function, the starting radius μ_0
 - Other methods

Algorithm

- 2 stages

- Pruning

- Exploration of the other branches

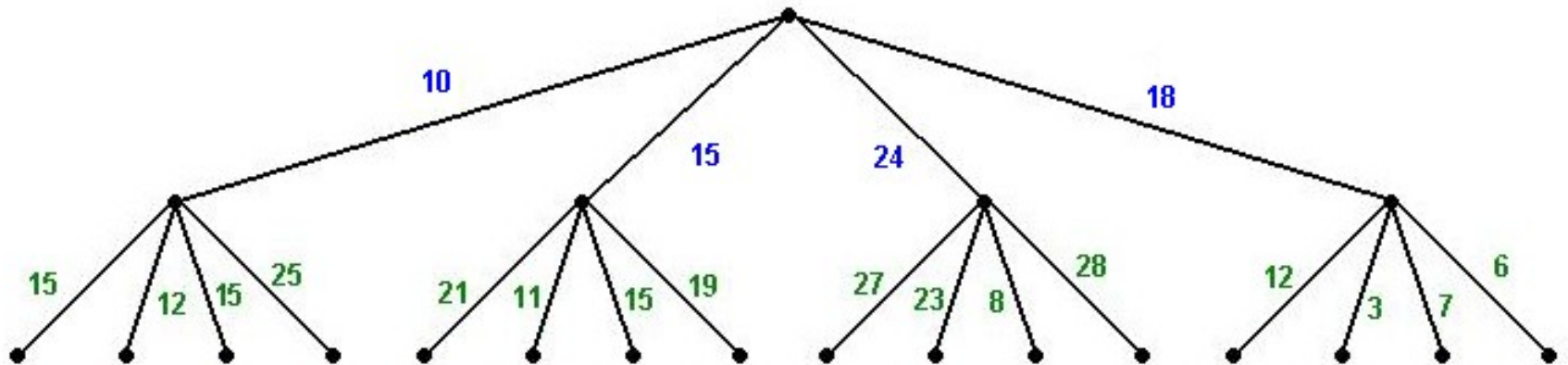
- Each branch which will certainly give a higher μ_0 is pruning out

- If a leaf is reached with a smaller sum than μ_0 , μ_0 is updated with that new value

- The process continues until all branches have been explored or pruned out

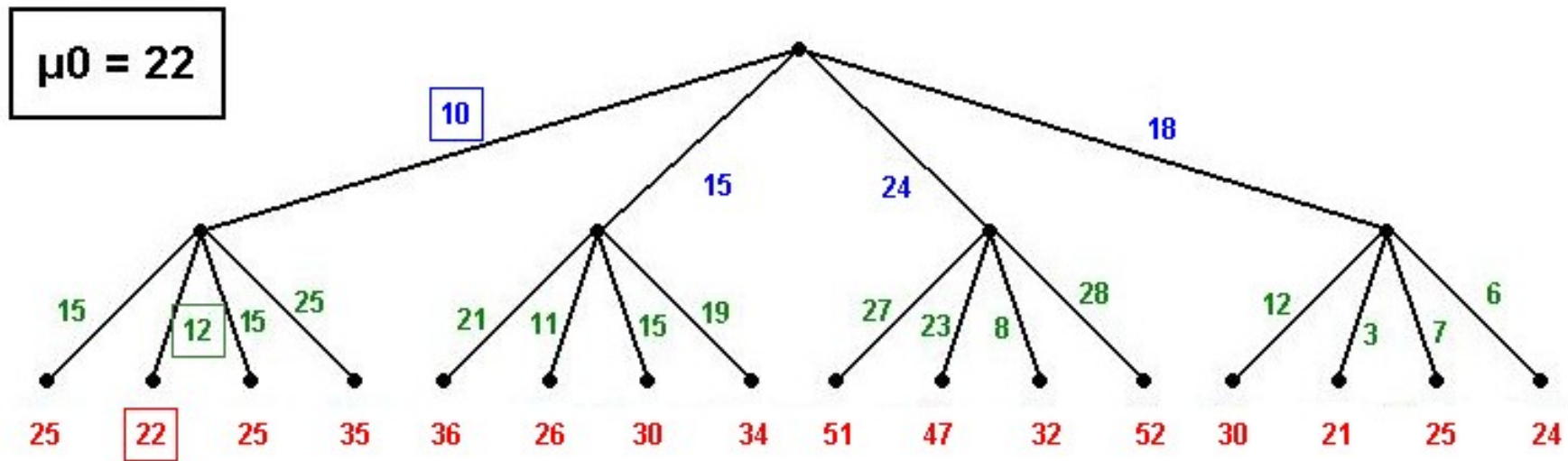
Exemple: $K=2, L=4$

□ Initialization



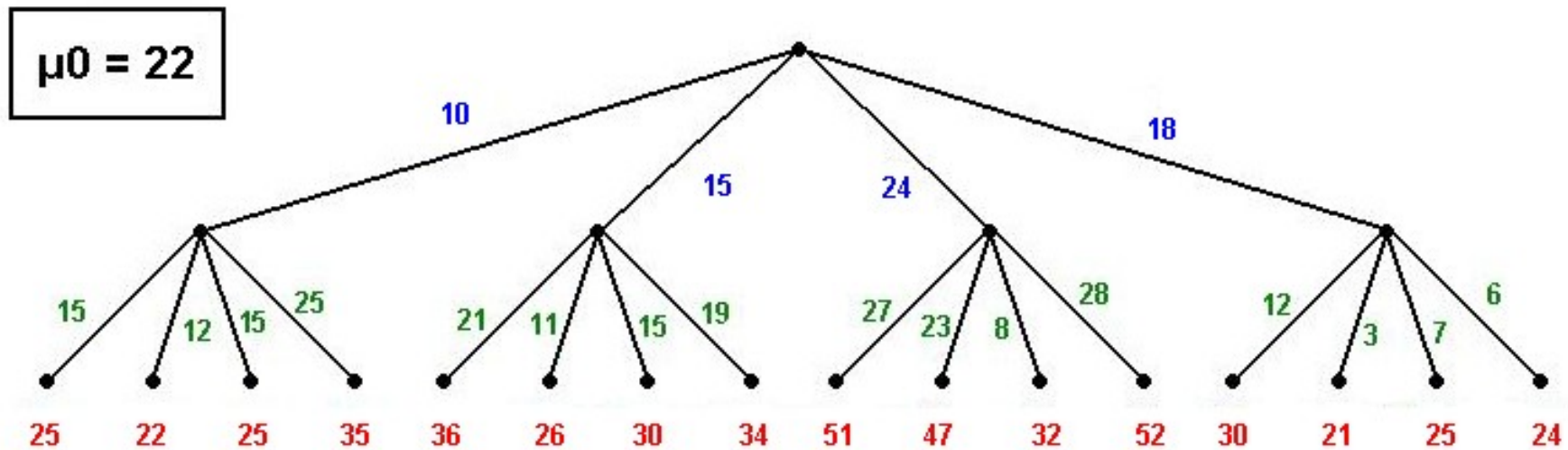
Exemple: $K=2$, $L=4$

Initialization



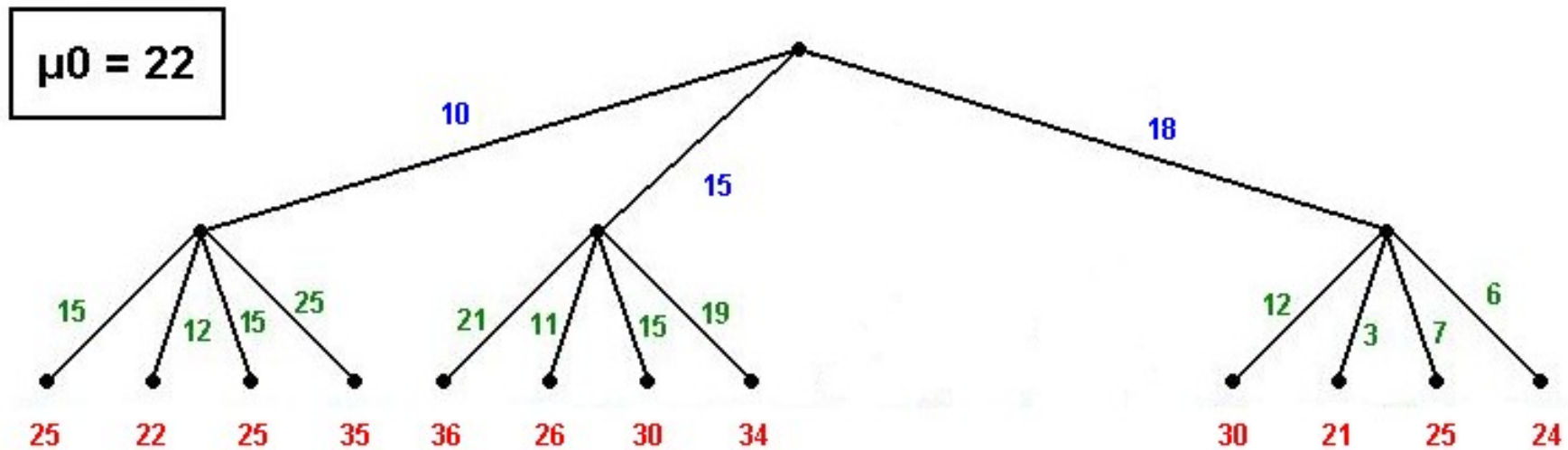
Exemple: $K=2, L=4$

□ Pruning



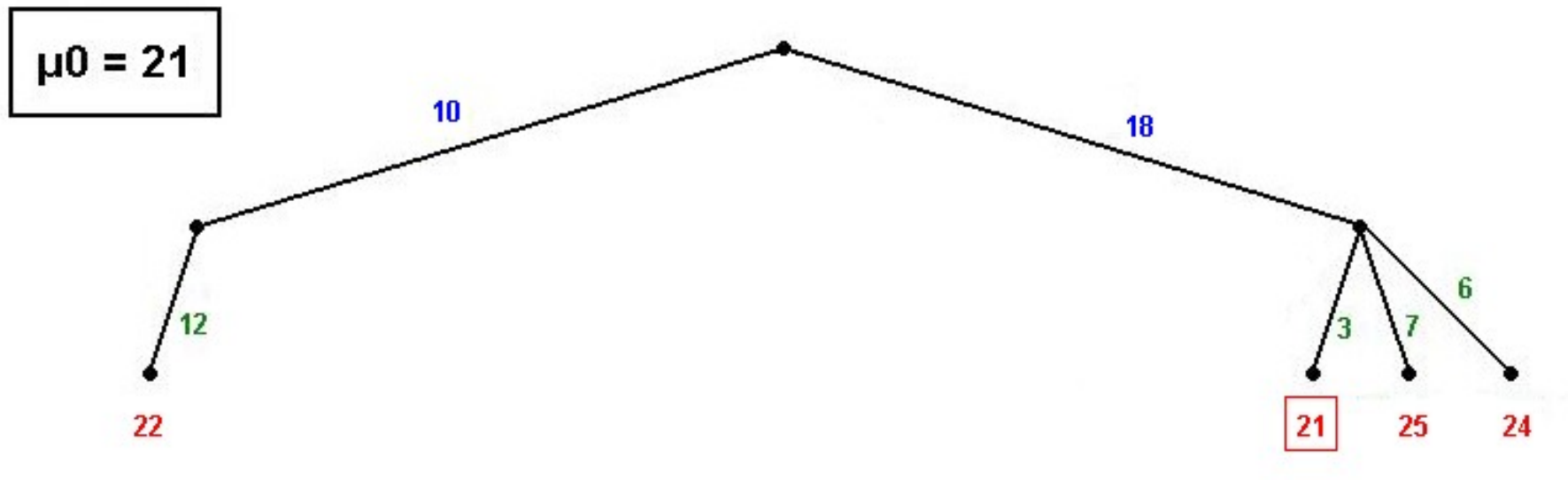
Exemple: $K=2, L=4$

□ Pruning (first level)



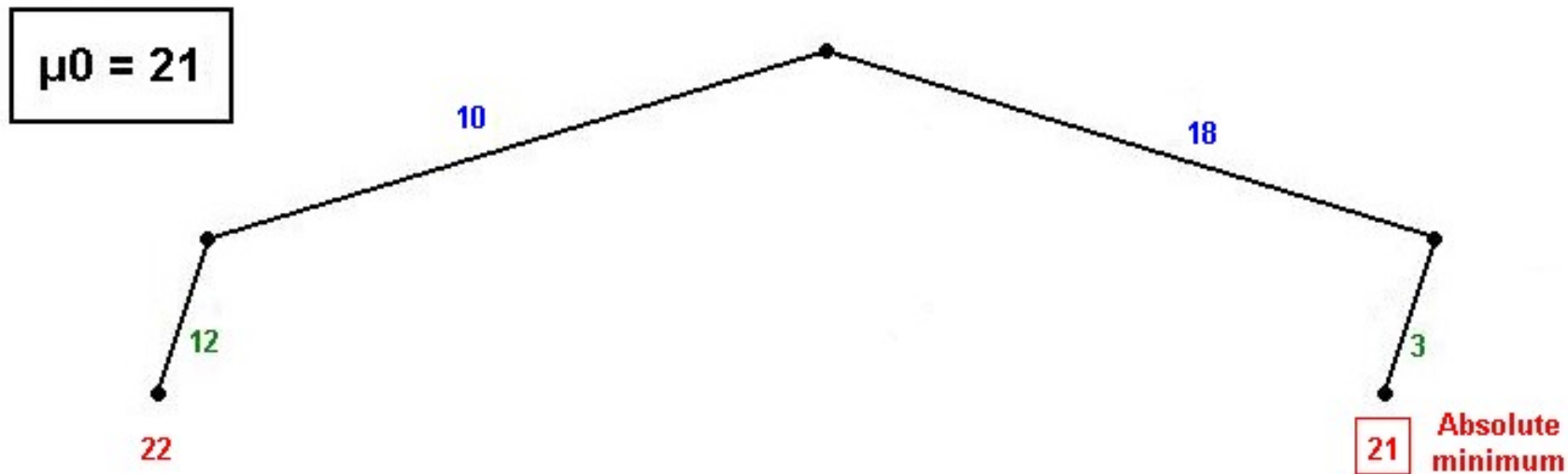
Exemple: $K=2$, $L=4$

□ Pruning (second level)



Exemple: $K=2, L=4$

□ Pruning (second level)



Conclusion

- The algorithm gives the absolute minimum of the cost function
- The complexity is smaller than a classic ML algorithm when the SNR grows as the pruning is more and more efficient
- The information of all the users is recovered in one time

Application to a MC-CSK system

Multiple Carrier Code Shift Keying

Reminder

- Spectral efficiency

$$\eta_s = R_b/B \text{ (bit.Hz}^{-1}\text{.s}^{-1}\text{)}$$

- Power efficiency

$$\eta_p = d^2/E_b$$

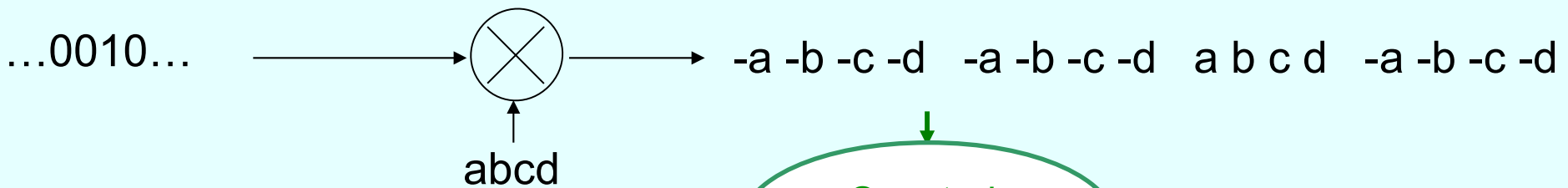
- Trade-off between η_s and η_p

MC-CSK system

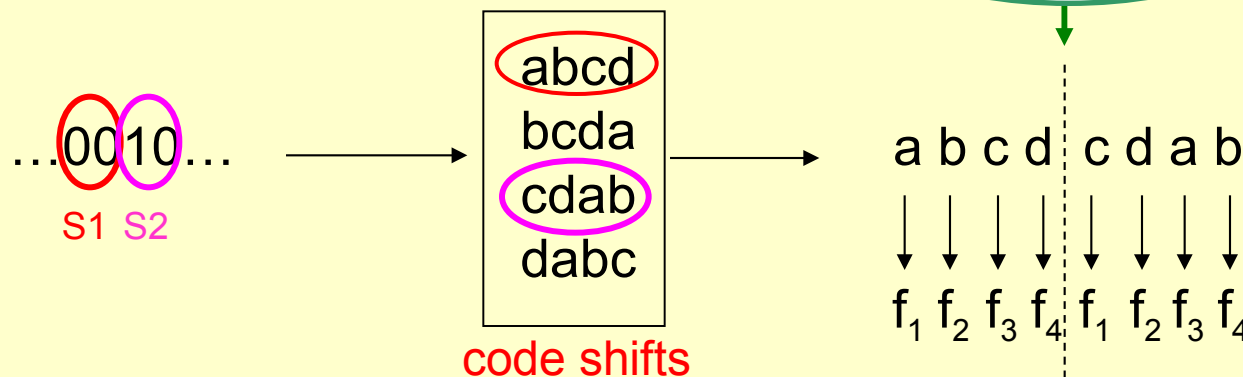
- Broadcast synchronous system
- K flows assimilated to K “users”
- Each user
 - L spreading sequences of length N (Gold sequences)
 - Coding $\log_2(L)$ bits
- Multi-user detection, not interference limited

MC-CSK system

CDM : code division multiplexing (N=4)



CSK : code shift keying (N=4, L=4)

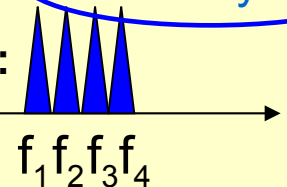


Spectral
efficiency $\times \log_2(L)$

Potentially good
power efficiency

Easier synchro

MC-CSK :



MC-CSK system

$\underline{\mathbf{C}}_1$ = matrix containing all the sequences of user 1

$\underline{\mathbf{C}} = [\underline{\mathbf{C}}_1 \ \underline{\mathbf{C}}_2 \ \dots \ \underline{\mathbf{C}}_K]$ = generation matrix

\mathbf{g}_1 vector of length L containing one '1' and $L-1$ '0'. The '1' points out the transmitted sequence for user 1

$\mathbf{g} = [\mathbf{g}_1 \ \mathbf{g}_2 \ \dots \ \mathbf{g}_K]^T$

Received signal: $\mathbf{z} = \underline{\mathbf{C}}.\mathbf{g} + \mathbf{w}_n$

MC-CSK system and Sphere Decoding

□ In reception

- Estimation of $\mathbf{g} = [\mathbf{g}_1 \ \mathbf{g}_2 \ \dots \ \mathbf{g}_K]^T$

- $\hat{\mathbf{g}} = \arg \min_{\mathbf{g}_1, \dots, \mathbf{g}_K} ||\mathbf{z} - \underline{\mathbf{C}}\mathbf{g}||_2^2$ (cost function)

- QR decomposition of $\underline{\mathbf{C}}$

 - $\underline{\mathbf{C}} = \underline{\mathbf{Q}}^* \underline{\mathbf{R}}$

 - $\underline{\mathbf{Q}}$ unitary, $\underline{\mathbf{R}}$ upper triangular

- $\hat{\mathbf{g}} = \arg \min_{\mathbf{g}_1, \dots, \mathbf{g}_K} ||\mathbf{y} - \underline{\mathbf{R}}\mathbf{g}||_2^2, \mathbf{y} = \underline{\mathbf{Q}}^{T*} \cdot \mathbf{z}$

MC-CSK system and Sphere Decoding

$$\underline{\mathbf{R}} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & r_{14} & \dots & r_{1,LK} \\ 0 & r_{22} & r_{23} & r_{24} & \dots & r_{2,LK} \\ 0 & 0 & r_{33} & r_{34} & \dots & r_{3,LK} \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & r_{LK-1,LK-1} & r_{LK-1,LK} \\ 0 & 0 & 0 & \dots & 0 & r_{LK,LK} \\ 0 & 0 & 0 & \dots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & 0 & 0 \end{bmatrix}$$

$$d = ||\mathbf{y} - \underline{\mathbf{R}}\mathbf{g}||^2 = h(\mathbf{g}_K) + h(\mathbf{g}_K, \mathbf{g}_{K-1}) + \dots + h(\mathbf{g}_K, \dots, \mathbf{g}_1)$$

MC-CSK system and Sphere Decoding

□ Initialization

- $\hat{\mathbf{g}}_K = \arg \min_{\mathbf{g}_K} h(\mathbf{g}_K)$

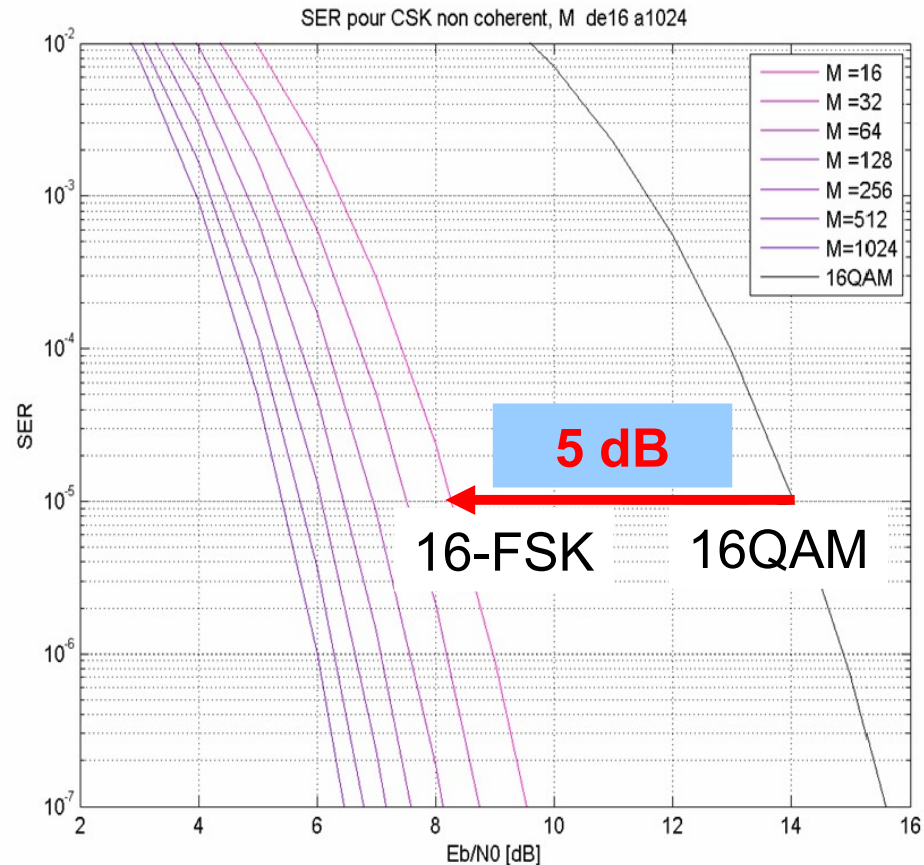
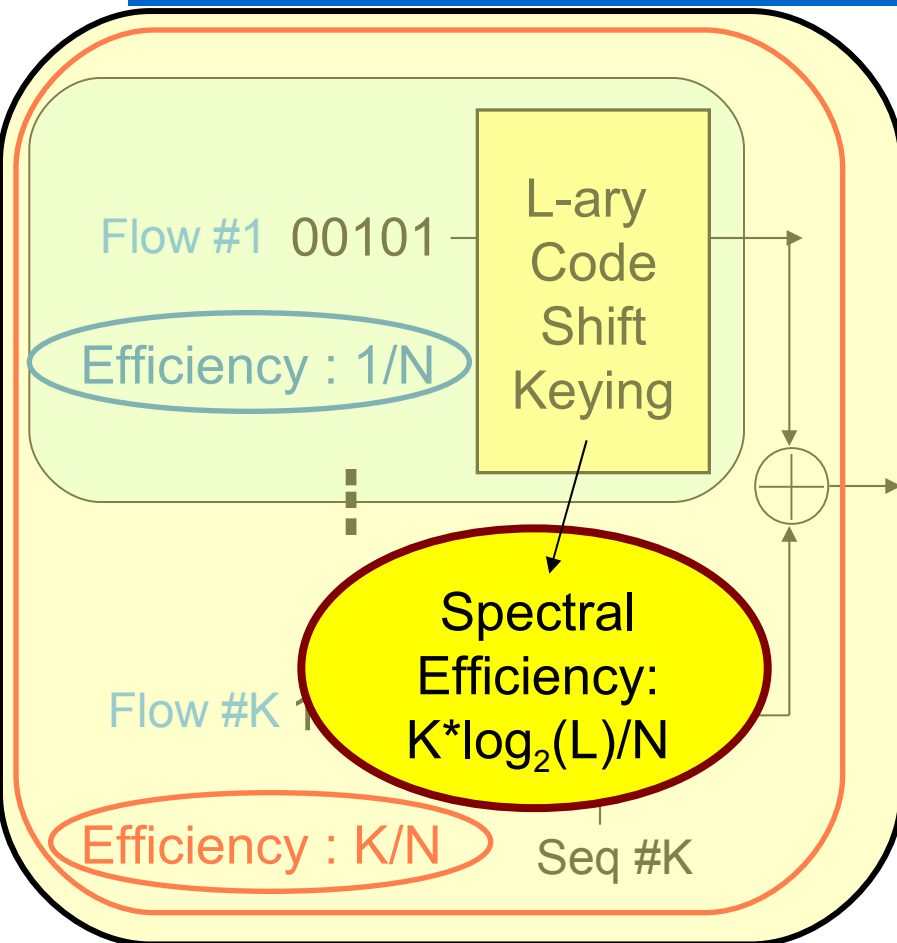
- $\hat{\mathbf{g}}_{K-1} = \arg \min_{\mathbf{g}_{K-1}} h(\hat{\mathbf{g}}_K, \mathbf{g}_{K-1})$

- $\hat{\mathbf{g}}_1 = \arg \min_{\mathbf{g}_1} h(\hat{\mathbf{g}}_K, \dots, \mathbf{g}_1)$

- $\mu_0 = h(\hat{\mathbf{g}}_K) + h(\hat{\mathbf{g}}_K, \hat{\mathbf{g}}_{K-1}) + \dots + h(\hat{\mathbf{g}}_K, \hat{\mathbf{g}}_{K-1}, \dots, \hat{\mathbf{g}}_1)$

□ Pruning

Why using MC-CSK and Sphere Decoding?

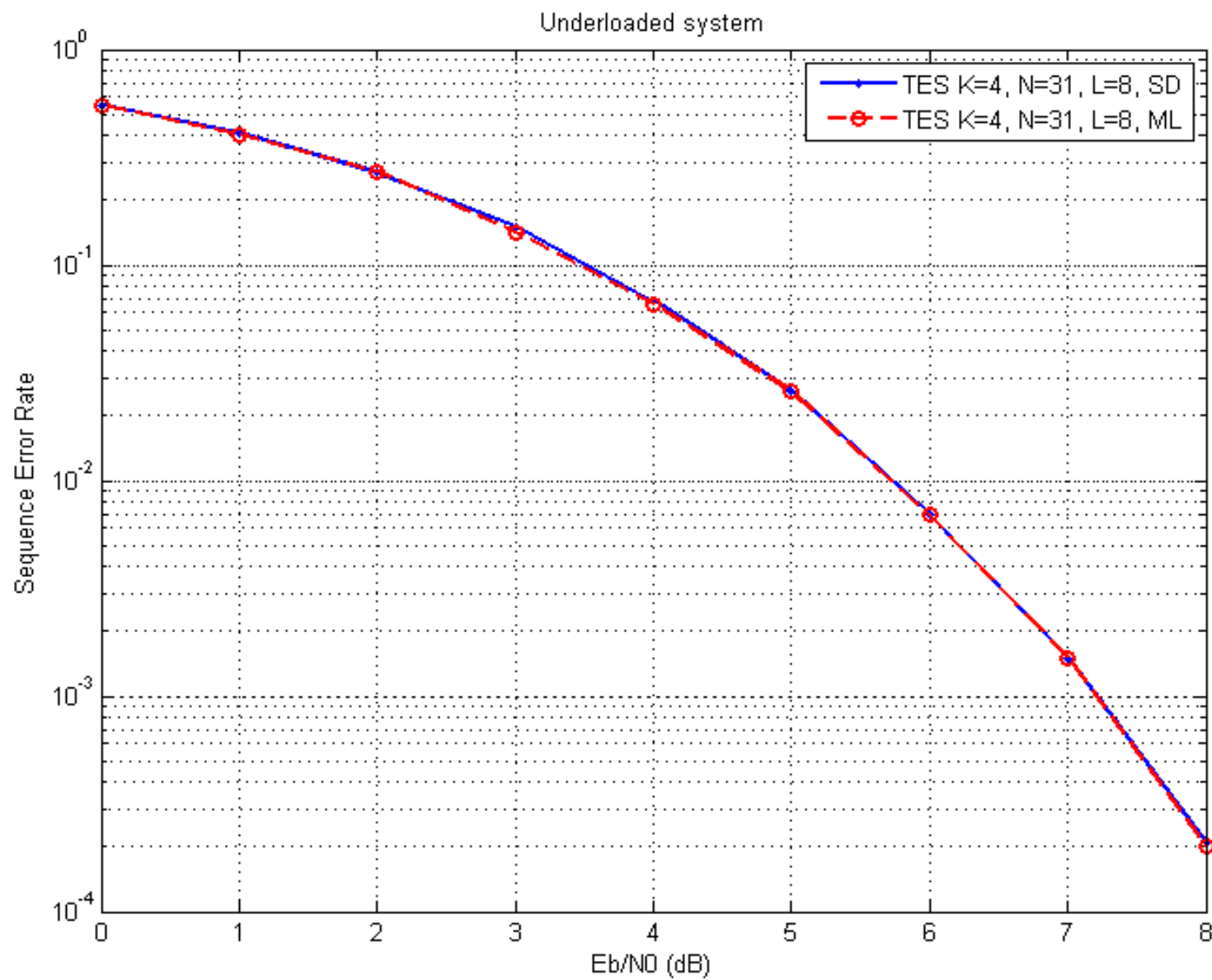


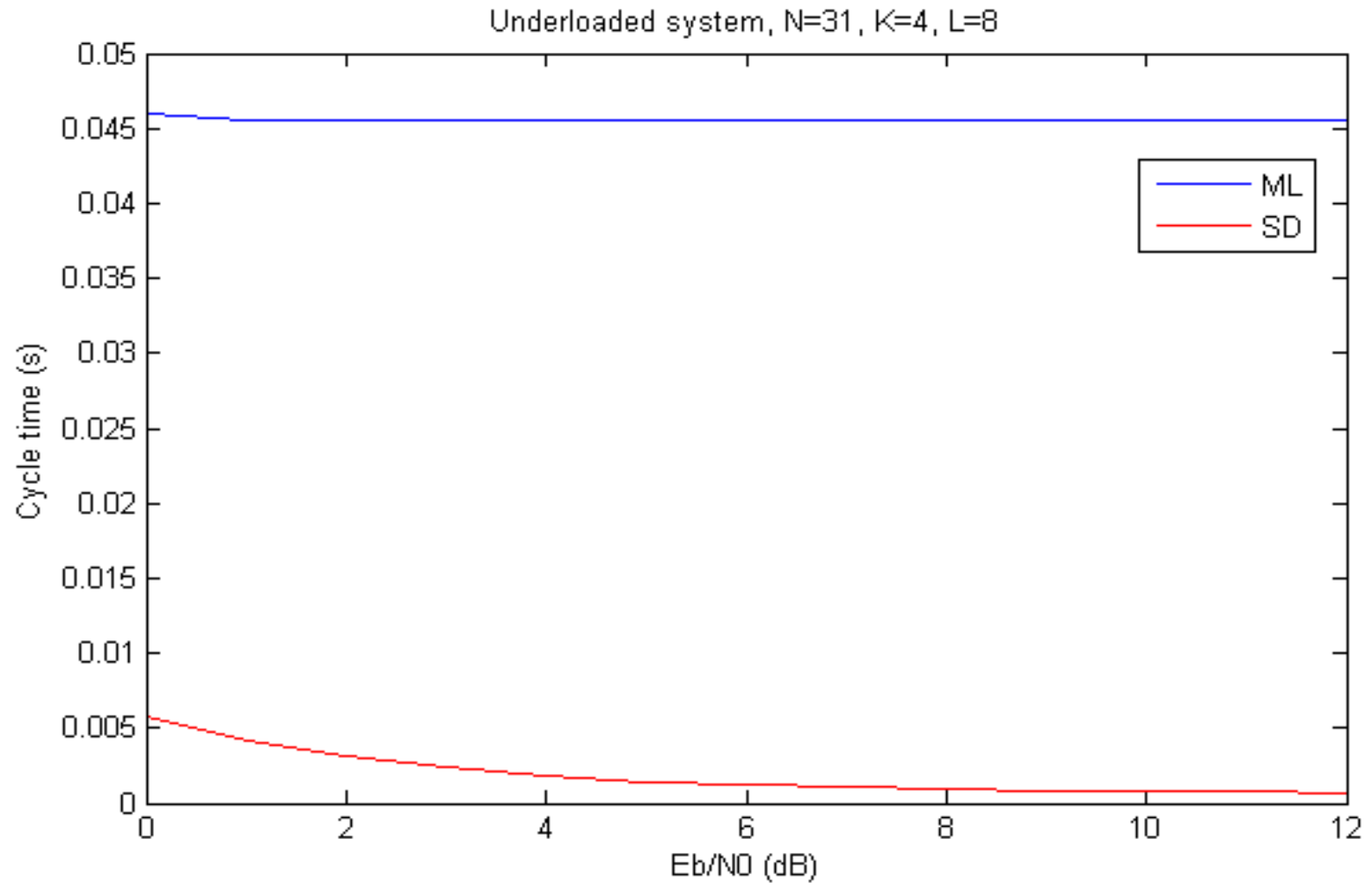
IF orthogonality of sequences & permutations
 → MC-CSK power efficiency
 = FSK power efficiency !

Why using MC-CSK and Sphere Decoding?

- ❑ System not interference limited
- ❑ Improving in the same time spectral and power efficiency
- ❑ Optimal decoding with lower complexity
- ❑ Scalable to (small) overloaded systems

Results





Limitations

- For higher loads ($>15\%$), the complexity and time simulation are too important
- Need of another decoder to reach interesting spectral efficiency
- LASSO

Brief presentation of the LASSO

- Least Absolute Shrinkage and Selection Operator – (Tibshirani, 1994)
- Working on the hypothesis of the sparsity of vector **g**
- Using a penalty on the L1 norm and the residual sum of squares

$$\min || \mathbf{y} - \mathbf{Cg} ||_2^2 \text{ s.t. } || \mathbf{g} ||_1^2 \leq t$$

Thank you for your attention

Questions?