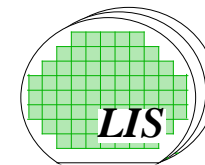


Adaptive Frequency Domain Equalization for OFDM

(Orthogonal Frequency Division Multiplex)

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outline

part 1: introduction

- **OFDM - basics**
- **fixed radio access scenario and channel models**

part 2: adaptive frequency domain equalization for OFDM

- **linear MIMO detectors for OFDM**
- **LMS- and RLS-algorithms**
- **modified RLS-algorithms (MRLS)**
- **differential coding**

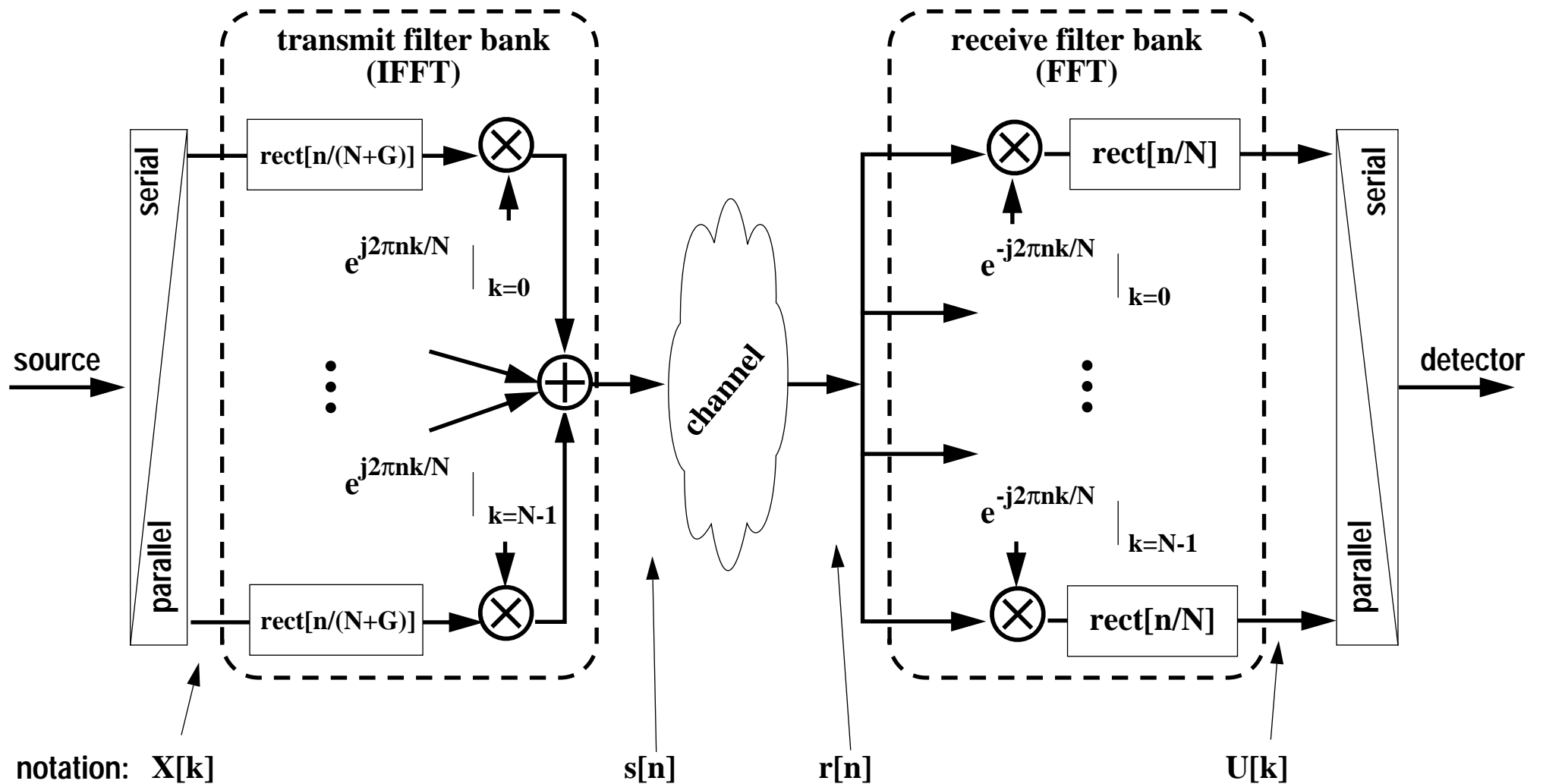
part 3: implementation

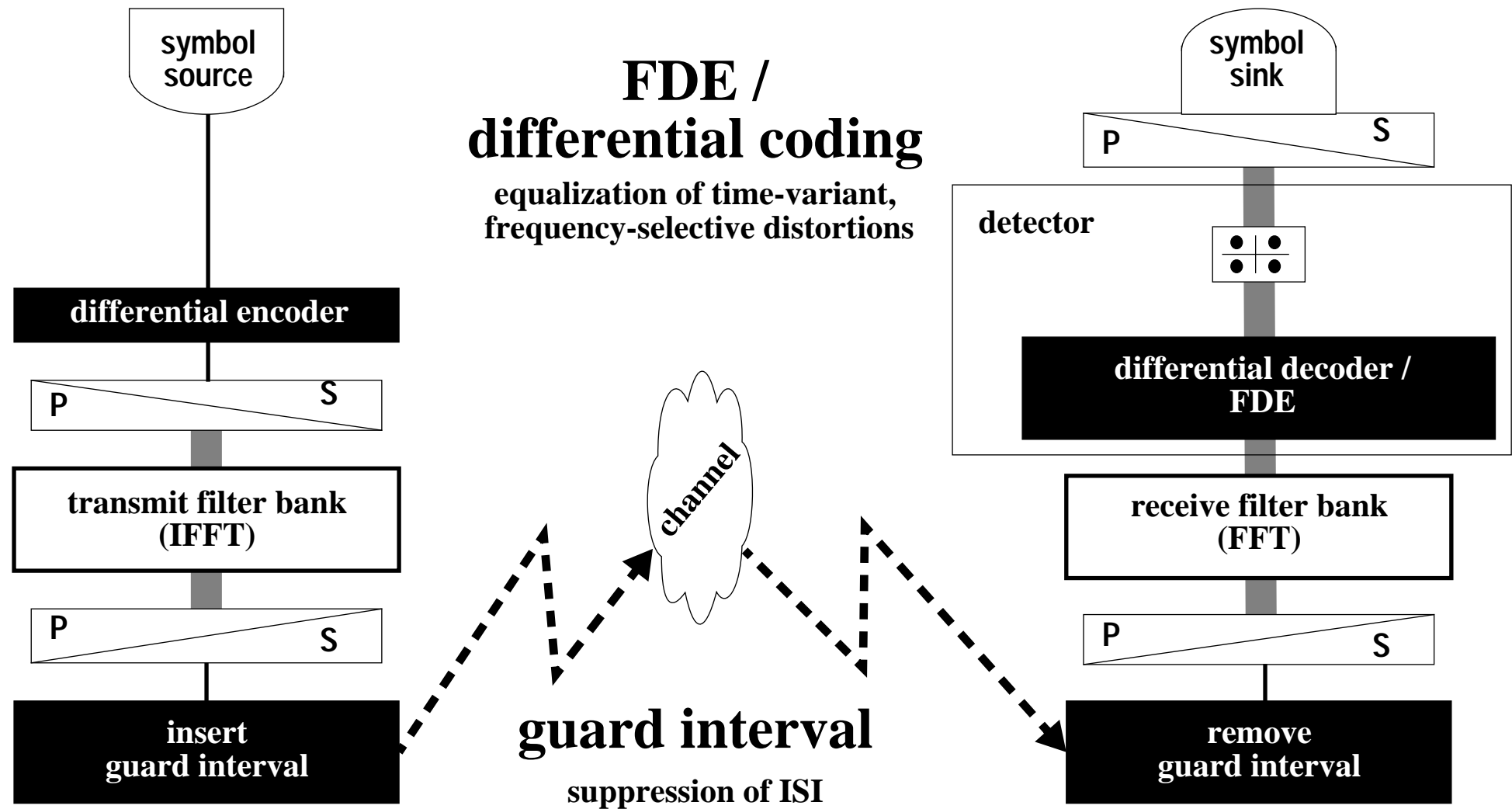
- **novel QRD-based square-root MRLS-algorithm**
- **complexity**

part 4: summary and outlook



Orthogonal Frequency Division Multiplex (OFDM)





(I)FFT (inverse) fast Fourier transform

P/S

parallel / serial conversion

ISI

inter-symbol interference

FDE

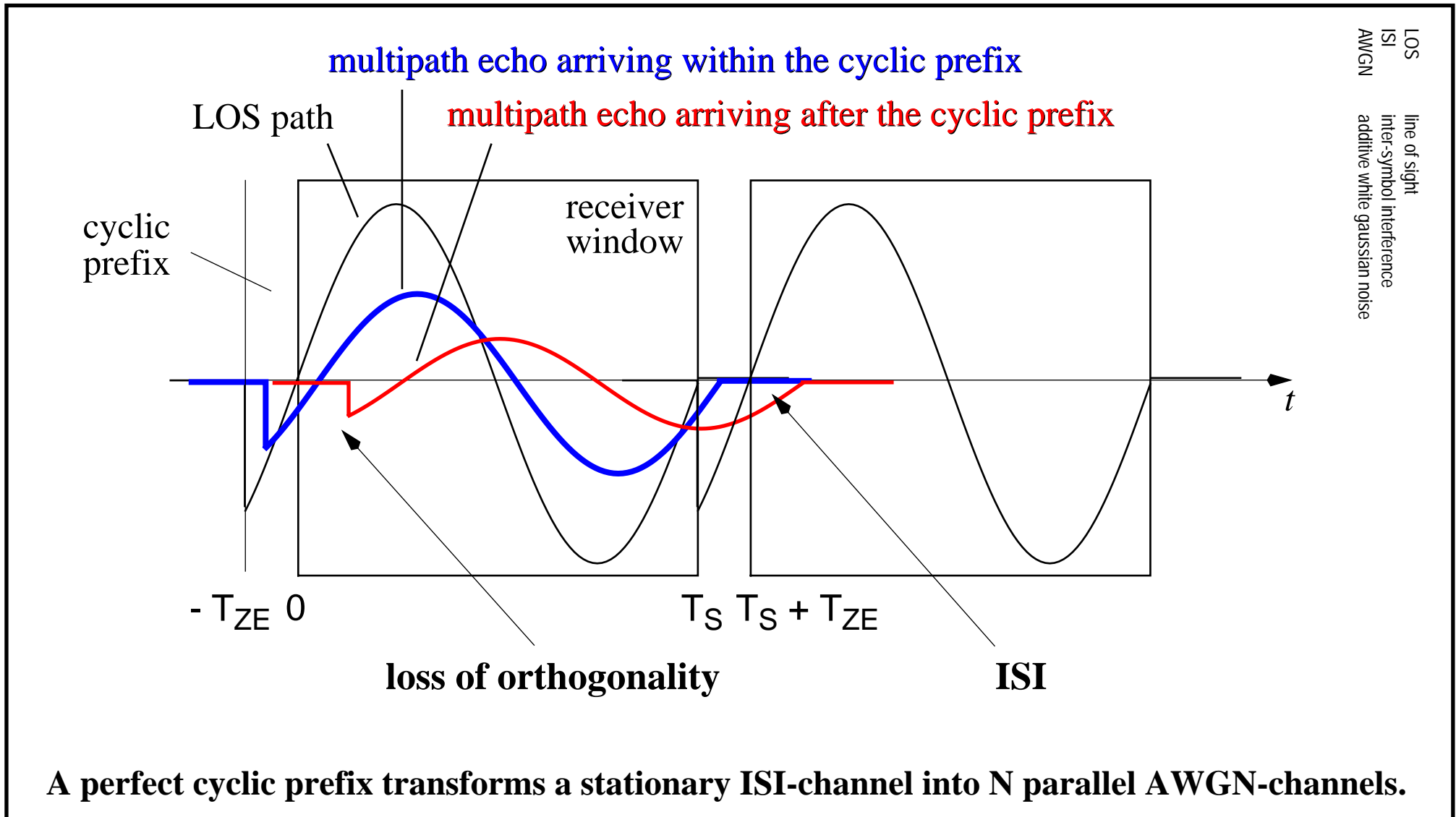
frequency domain equalization



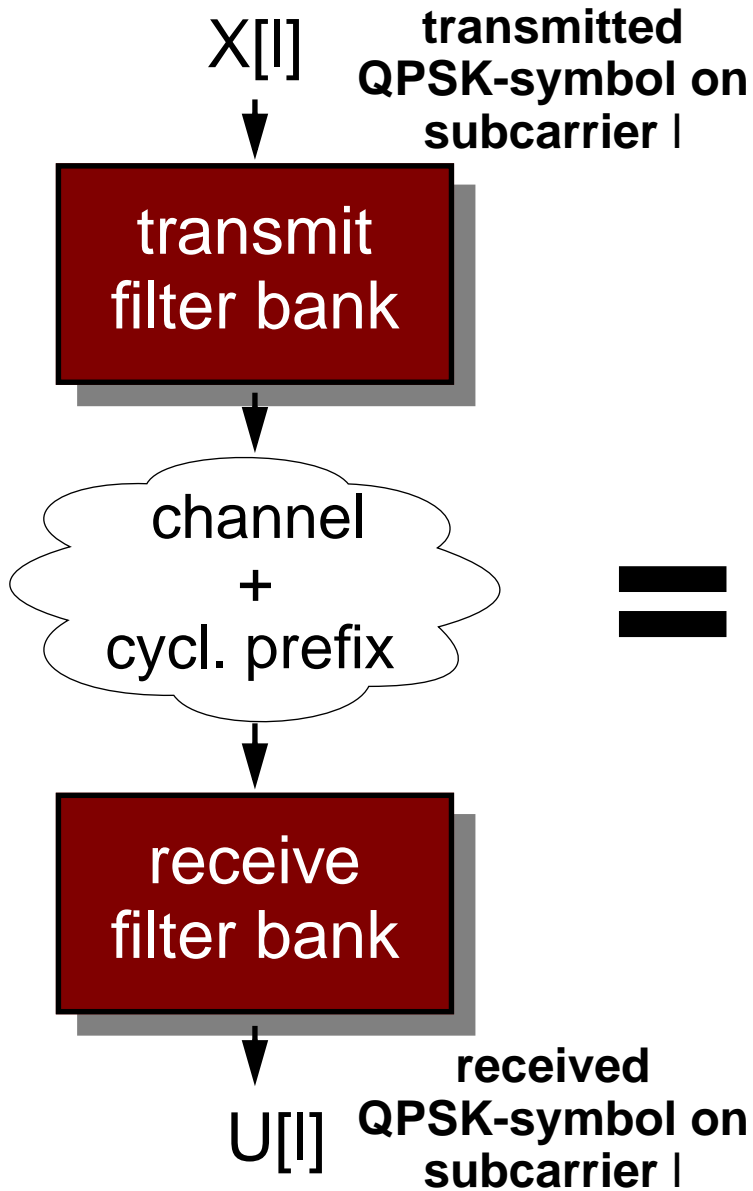
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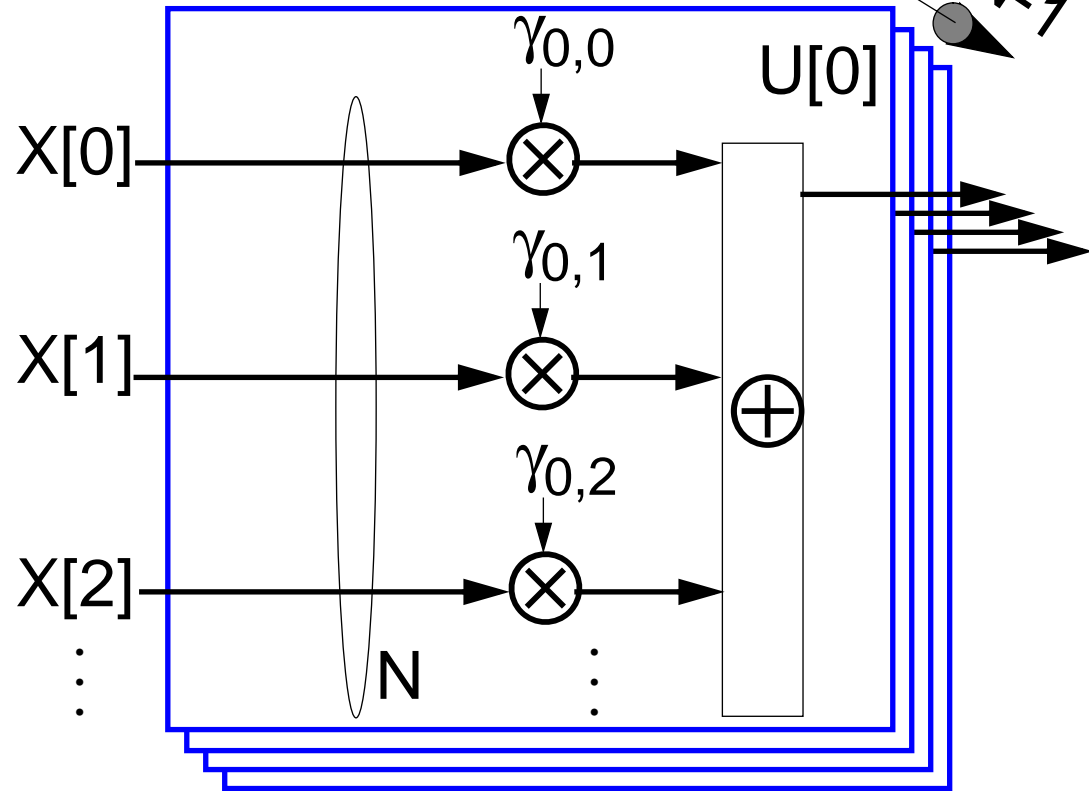


Multiple-Input Multiple-Output (MIMO) Channel Model - I



=

equivalent MIMO channel model



➤ no inter-symbol interference

QPSK
quaternary phase shift keying

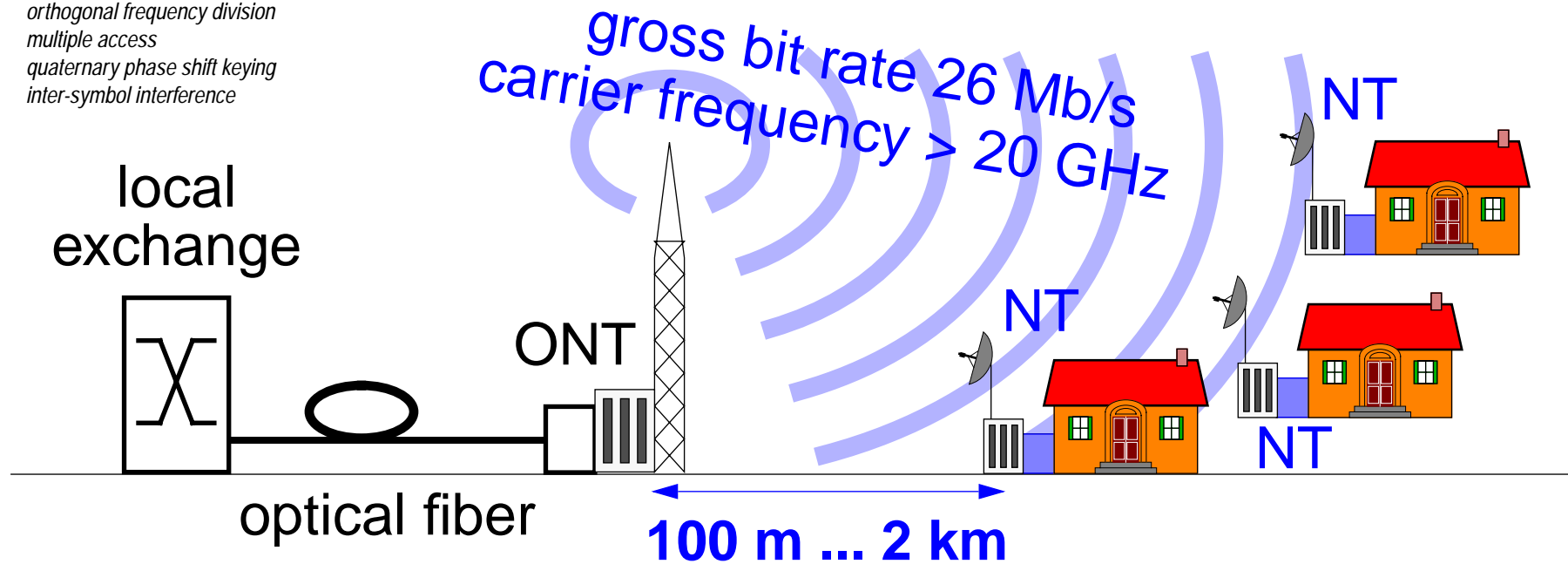


- **no inter-symbol interference (ISI)**
the transmission of a single OFDM-symbol can be considered

- **inter-carrier interference (ICI)**
 - ❑ **phase noise**
 - ❑ **synchronization errors**
 - ❑ **time-variant fading channels**
 - ❑ **...**

- **individual distortions on each subcarrier**

NT network termination
ONT optical network termination
OFDMA orthogonal frequency division multiple access
QPSK quaternary phase shift keying
ISI inter-symbol interference



OFDM / OFDMA proposal with $N = 512$ subcarriers and QPSK:

- short channel impulse responses with respect to the OFDM symbol duration
- negligible capacity loss introduced by perfect guard intervals
- all ISI are completely eliminated
- each OFDM-symbol is considered separately

worst case: omnidirectional antennas

- ❑ rician line of sight

- ❑ multipath propagation

 - 90 % of energy are received within 150 ns

 - echoes attenuated less than 15 dB are received within 200 ns

- ❑ reflections on moving cars are strongly attenuated

 - Rayleigh distributed Doppler components

 - 80 km/h -> maximum Doppler frequency 2 kHz @ 28 GHz

Sources: WALES, et al. (1993), GROND, et al. (1990): measurements in Bristol, UK

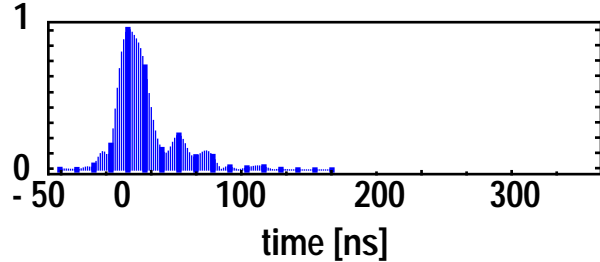
KARTHAUS, et al. (1997 - 1998): measurements within the *ATMmobil*¹ project

¹ supported by the German government

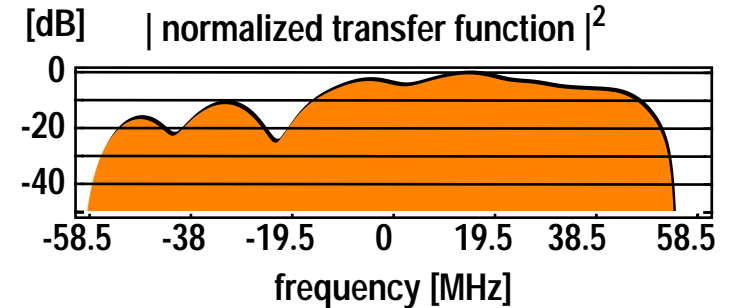
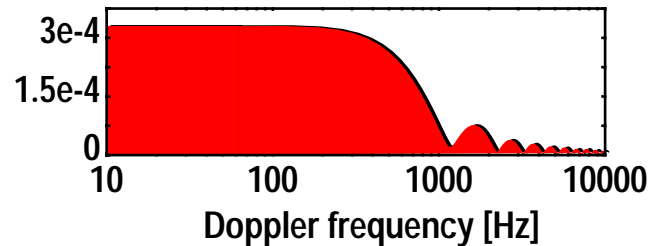


fading channel: omnidirectional antenna at the subscriber, sectorized antenna at the base station

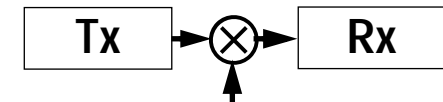
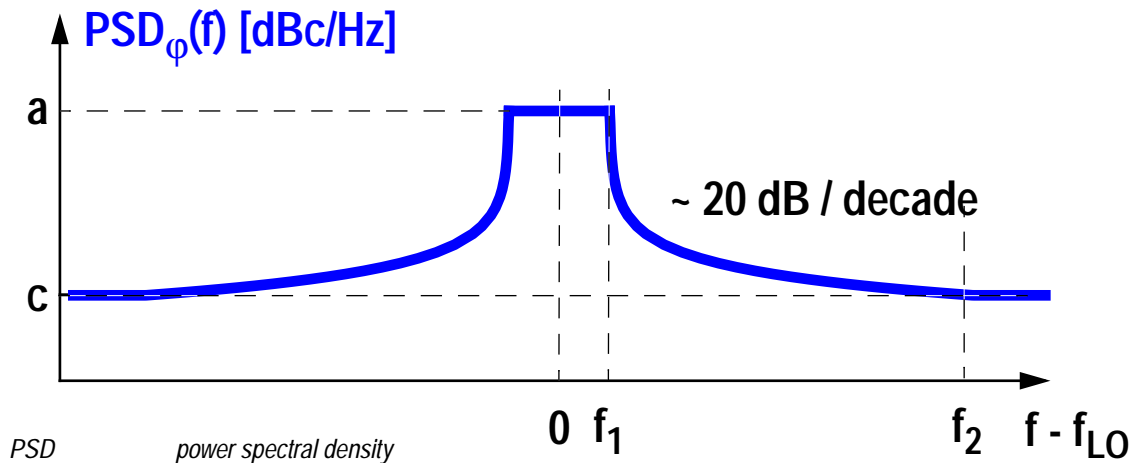
normalized impulse response



a) stationary
b) normalized Doppler spectrum



oscillator phase noise: $f_{LO} = 38$ GHz, $f_1 = 20$ kHz, $a = -65$ dBc/Hz, $c = -125$ dBc/Hz



$$p[n] = e^{j\varphi[n]} \approx 1 + j\varphi[n]$$

$\varphi[n]$ zero mean gaussian process
 $\varphi[n] \ll 1$ rad



degradations due to phase noise

➤ common phase error (CPE)

- ❑ identical for all subcarriers during one OFDM-symbol
- ❑ easy to compensate
- ❑ → 0 with increasing bandwidth of phase noise
- ❑ → 0 with increasing number of subcarriers

➤ inter-carrier interference (ICI)

- ❑ nearly gaussian for high numbers of subcarriers
- ❑ = 0 for stationary phase noise
- ❑ ICI-power increases with the number of subcarriers

OFDM transmission with cyclic prefix is optimum in ISI-channels in terms of Shannon's capacity, provided:

- ❑ the number of subcarriers tends to infinity**
- ❑ the transmit power is distributed among all subcarriers in an optimum way (loading algorithms)**

Time Division Multiple Access (TDMA) + Adaptive Decision Feedback Equalizer

Radio Channel:

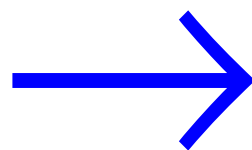
multipath delay: 20 ... 300 ns
coherence time: $T_C = 720 \mu\text{s}$

Uplink:

minimal data-rate: 32 kb/s
separation of two ATM-cells: 12 ms $\gg T_C$

**Fast converging algorithms
have to be used (RLS)**

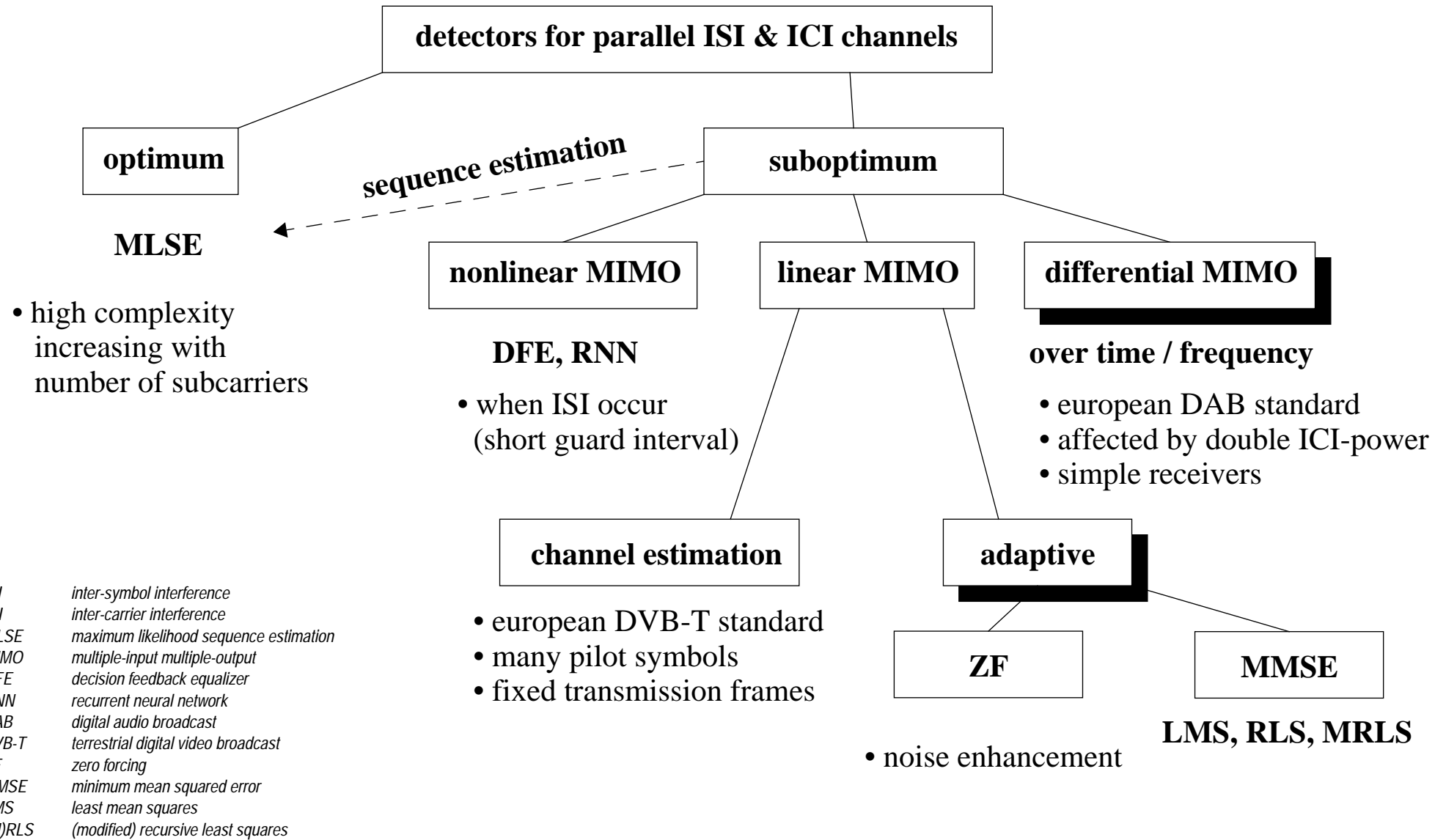
→ high complexity
(100,000 - 200,000 gates with a
0.35 μm standard cell library)



**Does
multicarrier transmission
reduce the
implementation cost**



Detectors for OFDM - Some Concepts Proposed



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Since we want to analyze **adaptive equalization** ...

... we express the time-variance of the channel coefficients by means of

the time-variance of the optimum equalizer coefficients !

optimum coefficient vector
at time m

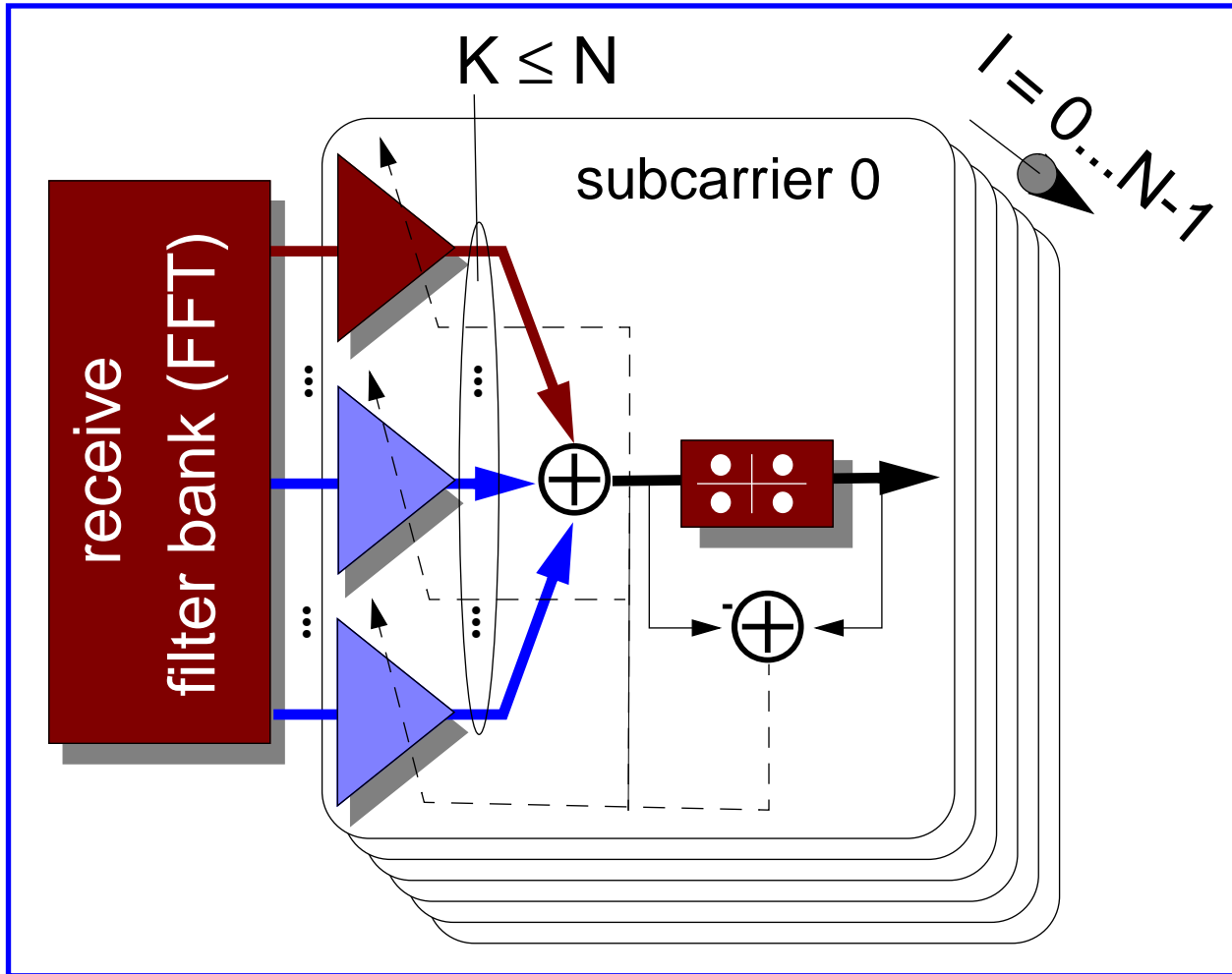
process noise vector

$$\hat{\mathbf{c}}_{m+1} = \hat{\mathbf{c}}_m + \mathbf{v}_{P,m}$$

dynamic first order Markoff model

MIMO multiple-input multiple-output





- linear equalization & ICI-compensation
- decoupling
equalizer coefficients of each subcarrier can be adapted independently
- parallel multiple-input single-output (MISO) structures
- ideal guard interval
→ 1 complex valued coefficient per branch

- ICI of adjacent subcarriers dominate
- $K \ll N$ branches per MISO-equalizer

FFT fast Fourier transform
 ICI inter-carrier interference
 DFE decision feedback equalizer
 MIMO multiple-input multiple-output

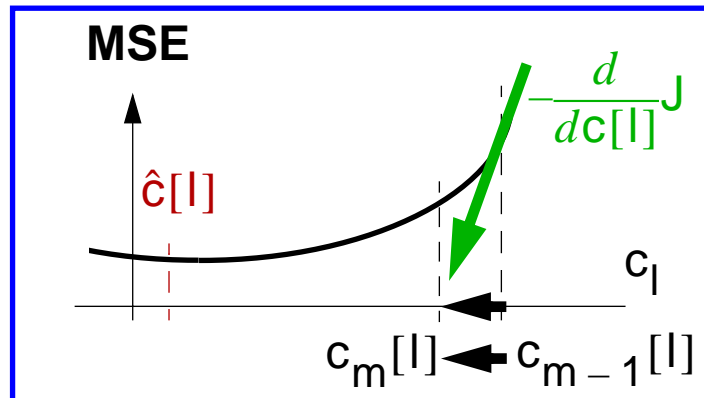


least mean squares (LMS) algorithm

cost function

mean squared error (MSE) on all subcarriers l

$$J = \sum_{l=0}^{N-1} E \left\{ |e_m[l]|^2 \right\}$$



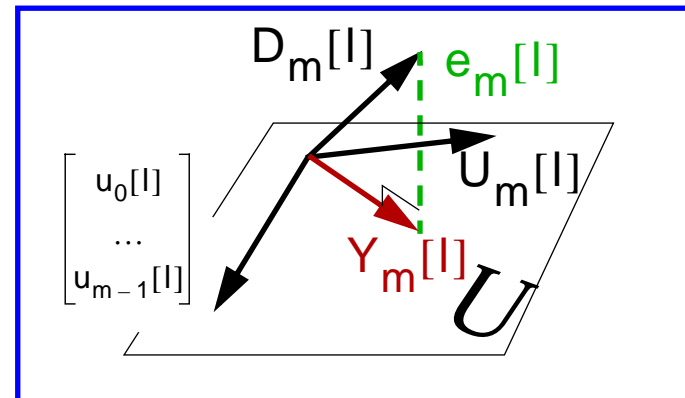
low complexity ...
but slow convergence

recursive least squares (RLS) algorithm

cost function

sum of all squared errors observed at time $0 \dots m$ on all subcarriers l

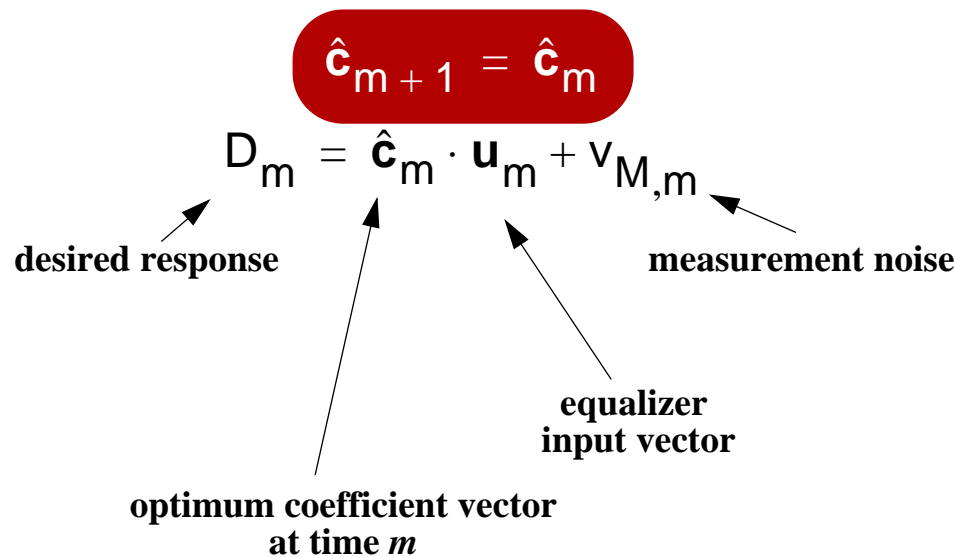
$$J = \sum_{l=0}^{N-1} \sum_{i=0}^m \lambda^{m-i} \cdot |e_m[l]|^2$$



fast convergence ...
but high complexity & numerical problems

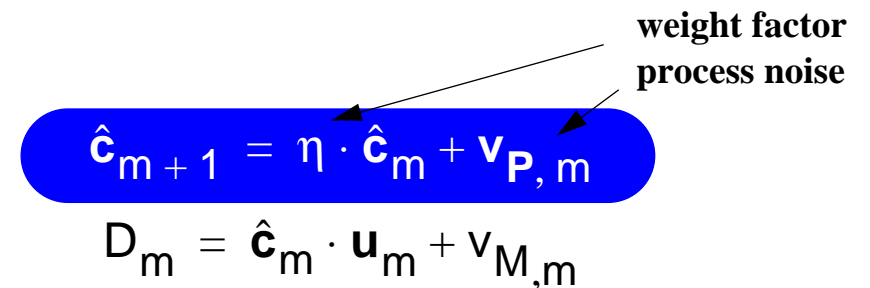
the RLS-algorithm is the special case of the Kalman filter for a **stationary state-space model !**

➤ suboptimum tracking properties



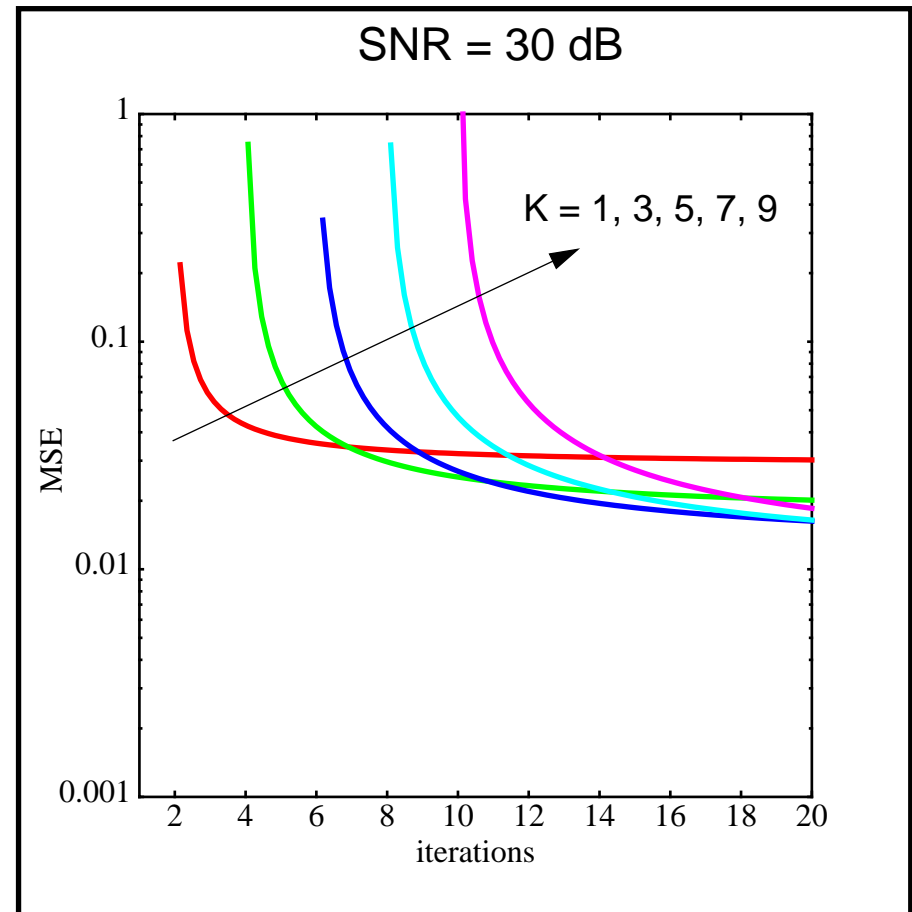
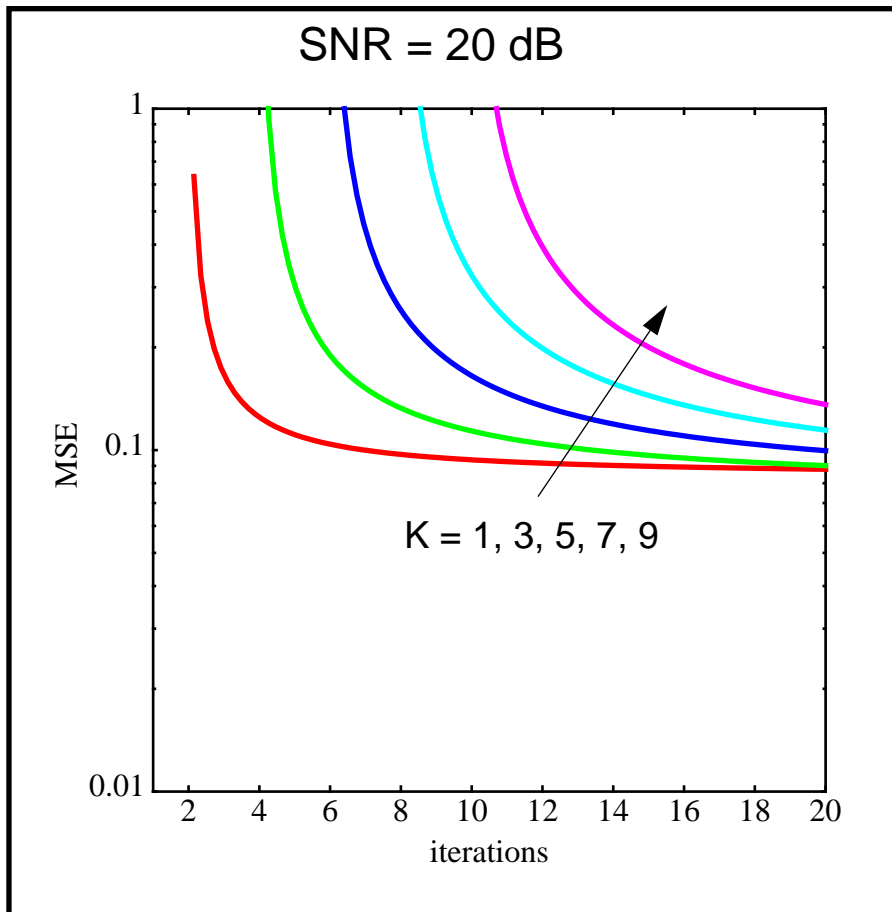
➤ design modified RLS-algorithms with better tracking properties from a **dynamic state-space model** (HAYKIN, et al. 1995, 1997)

first order Markoff model



➤ invoke the modification automatically only when the transmission channel is essentially changing (PARK & JUN, 1992 / JIANG & COOK, 1992)

Convergence of the MISO RLS Algorithm (Theory)



SNR signal to noise ratio
 MSE mean squared error
 RLS recursive least squares
 MIMO multiple-input multiple-output

- theoretical convergence in a stationary equivalent MIMO channel
- stationary frequency selective ICI channel



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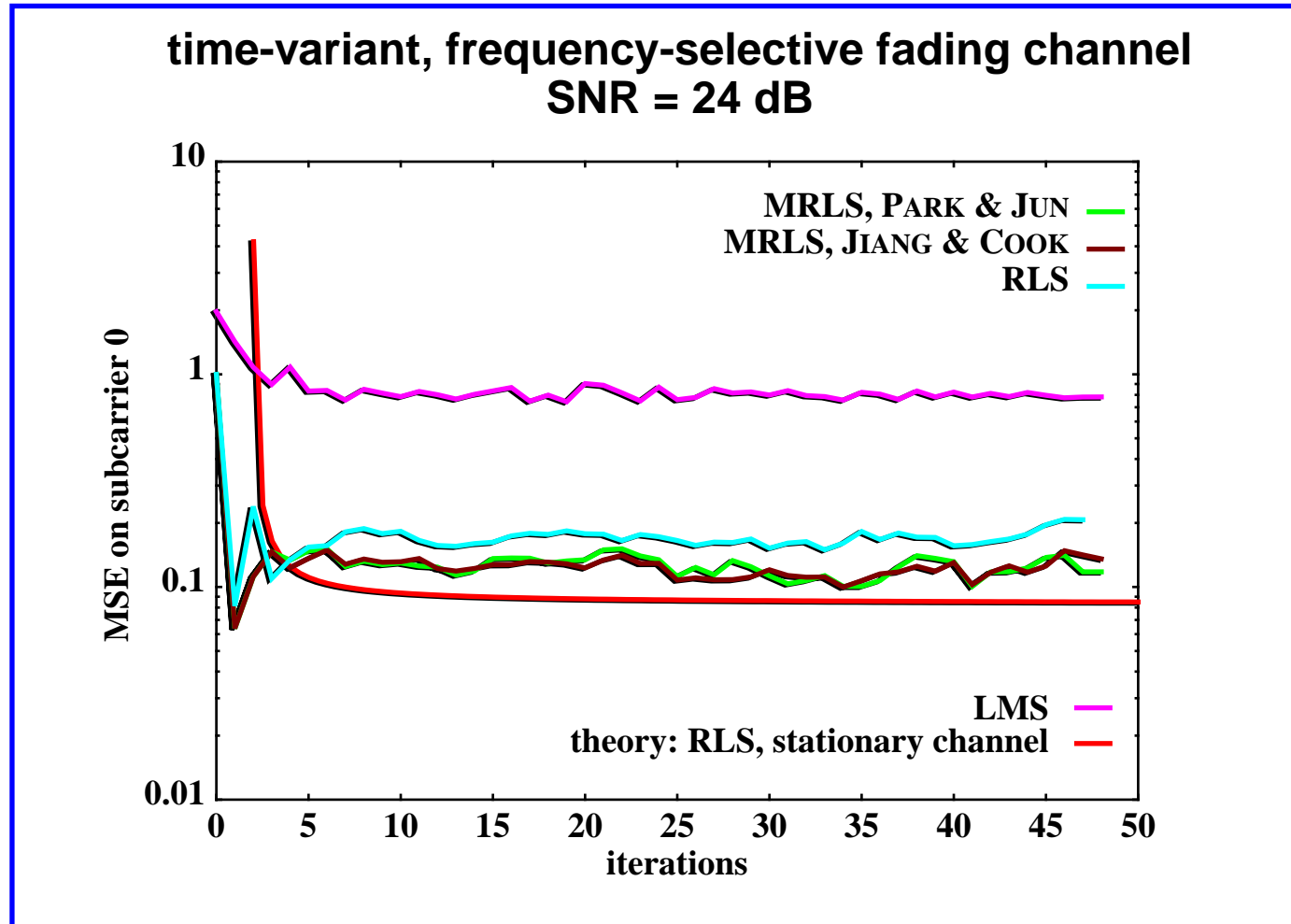
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Convergence of the MISO Equalizer - Simulation I

ICI negligible vs. AWGN

$\gg K = 1$



MSE
MISO
AWGN
ICI
mean squared error
multiple-input single-output
additive white gaussian noise
inter-carrier interference
SNR
LMS
(M)RLS
signal to noise ratio
least mean squares
(modified) recursive least squares

Kalman filter based modifications improve the tracking properties of the RLS algorithm.

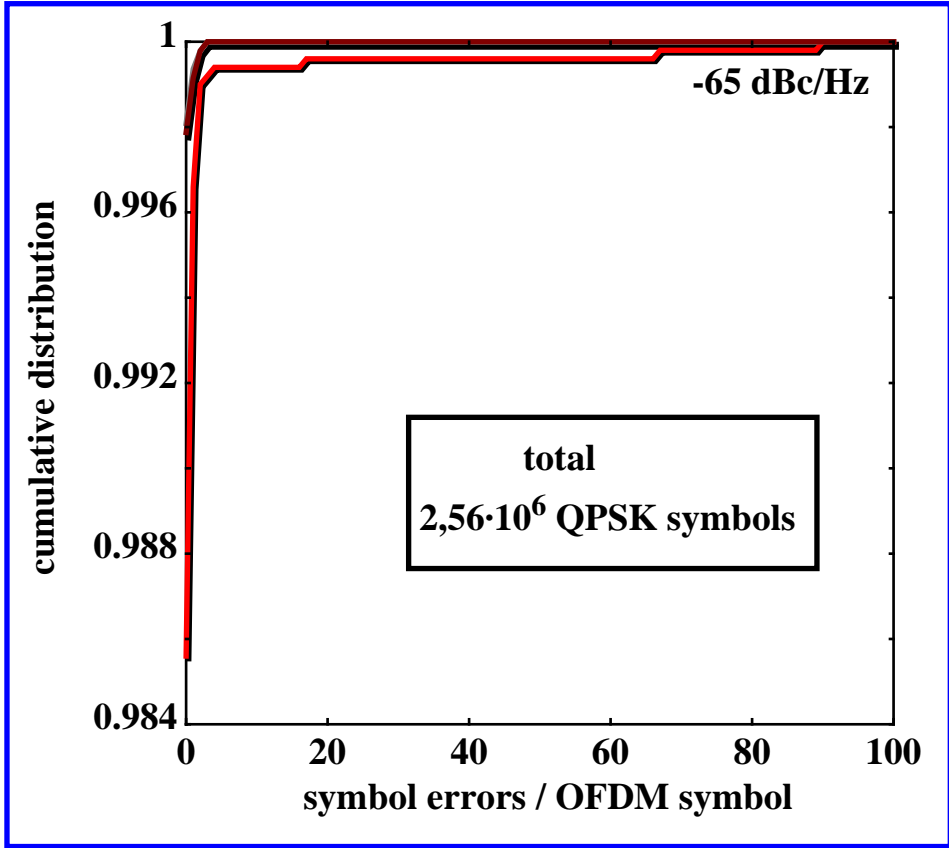
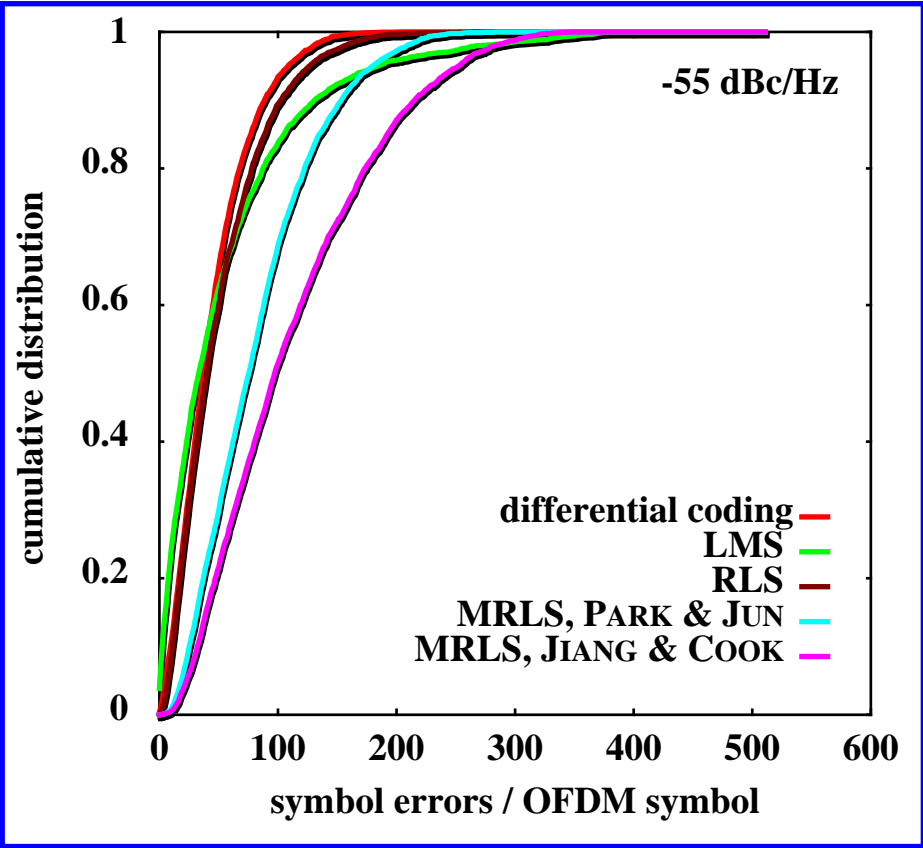


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stationary frequency-selective fading channel + phase noise



symbol error rate

d.c.	LMS	RLS	P&J	J&C
0.089	0.10	0.098	0.16	0.22

symbol error rate

d.c.	LMS	RLS	P&J	J&C
$1.0 \cdot 10^{-4}$	$1.2 \cdot 10^{-5}$	$5.4 \cdot 10^{-6}$	$3.8 \cdot 10^{-5}$	$6.2 \cdot 10^{-6}$



- **The capacity loss due to ideal cyclically extended guard intervals is negligible in fixed radio access scenarios.**
- **Phase noise introduces severe ICI.**
- **Linear adaptive FDE in combination with ideal guard intervals provides more flexibility than coherent channel estimation and is less sensitive with respect to ICI than differential coding.**
- **Modified RLS algorithms with better tracking capabilities can be designed from dynamic state-space models.**
- **Neither the RLS, LMS or MRLS algorithms, nor differential coding, can mitigate the requirements on the purity of oscillators in OFDM-systems significantly.**



RLS equalizers in their conventional form have several drawbacks:

high complexity

- $O(N^2)$ multiplications and additions, and $O(N)$ divisions per iteration
- fast RLS algorithms can not be applied to OFDM due to independent input vectors

numerical problems

- the inverse correlation matrix is computed as the difference of 2 positive definite matrices and may not remain meaningful in a finite precision environment

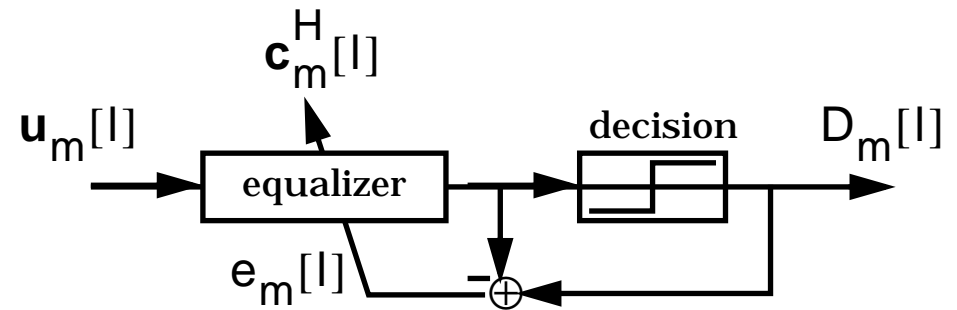
large wordlengths required

- the RLS algorithm operates on the square of the input vectors

These drawbacks can be tackled by square-root RLS algorithms !

applying the QR decomposition to the matrix of input vectors

$$\mathbf{U}_m = \begin{bmatrix} \sqrt{\lambda}^n \mathbf{u}_0 & \sqrt{\lambda}^{n-1} \mathbf{u}_1 & \dots & \mathbf{u}_m \end{bmatrix}^H$$



QR decomposition $\mathbf{Q}_m \mathbf{U}_m = \begin{bmatrix} \mathbf{R}_m \\ \mathbf{0} \end{bmatrix}$
(Q ... unitary, R ... upper triangular matrix)

$$\mathbf{D}_m = \begin{bmatrix} \sqrt{\lambda}^n D_0 \\ \sqrt{\lambda}^{n-1} D_1 \\ \dots \\ D_m \end{bmatrix}$$

$$\mathbf{e}_m = \begin{bmatrix} \sqrt{\lambda}^n e_0 \\ \sqrt{\lambda}^{n-1} e_1 \\ \dots \\ e_m \end{bmatrix}$$

multiplying the error vector by an unitary matrix does not change the cost function $|\mathbf{e}_m|^2$

$$\mathbf{e}_m = \mathbf{D}_m - \mathbf{U}_m \hat{\mathbf{c}}_m$$

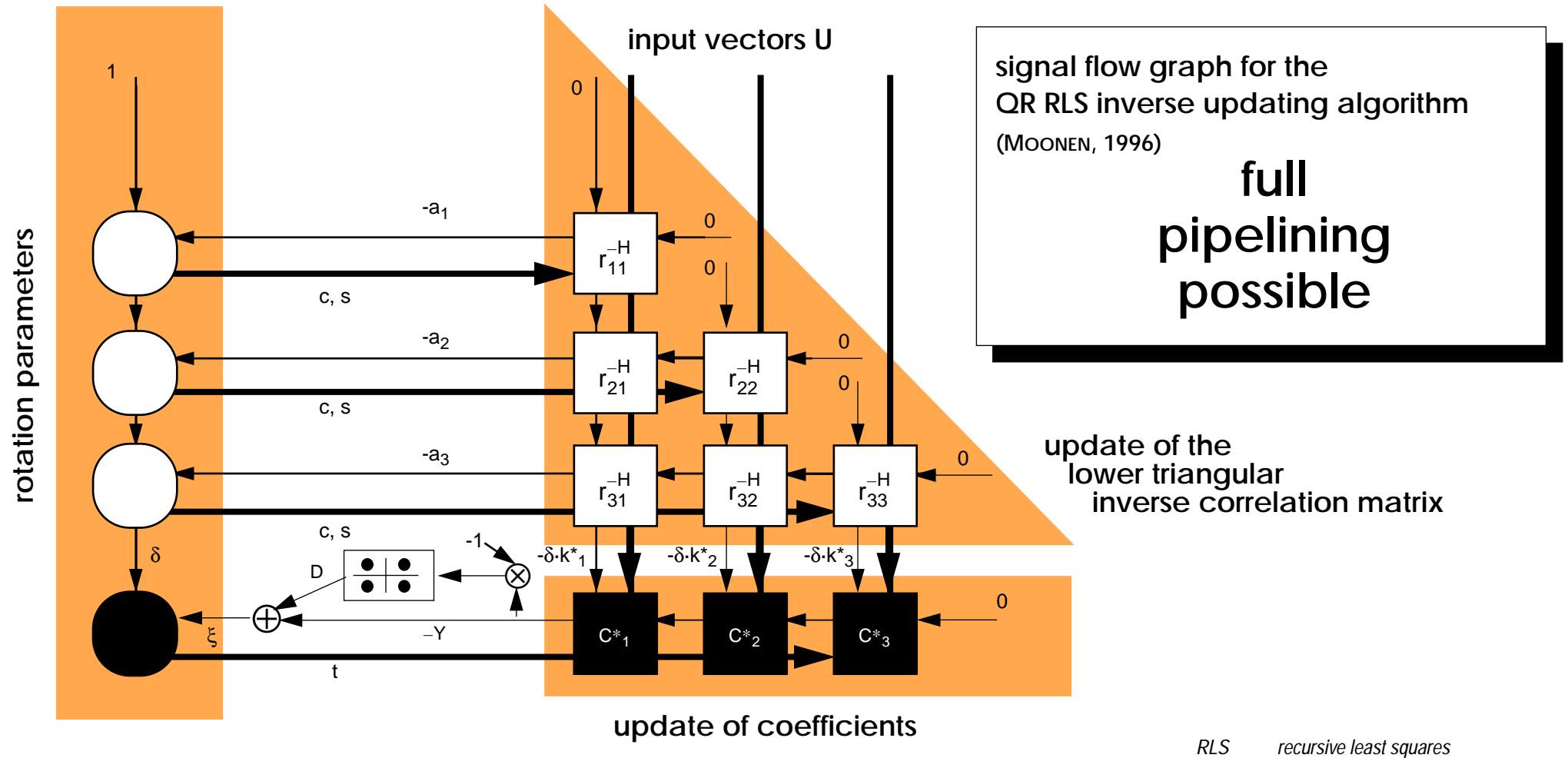
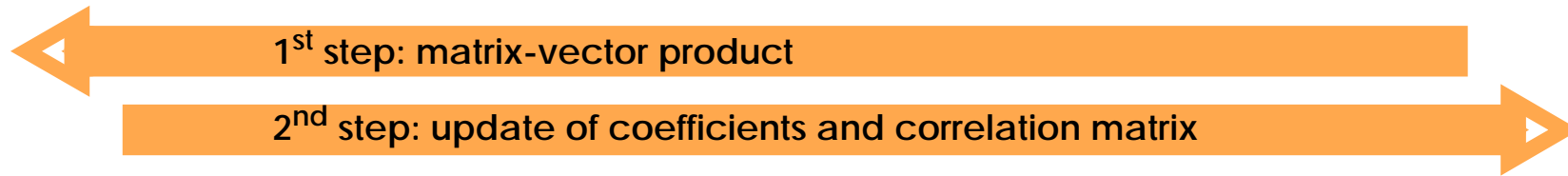
$$\mathbf{Q}_m \mathbf{e}_m = \mathbf{Q}_m \mathbf{D}_m - \begin{bmatrix} \mathbf{R}_m \\ \mathbf{0} \end{bmatrix} \hat{\mathbf{c}}_m$$

$$\hat{\mathbf{c}}_m = \mathbf{R}_m^{-1} \mathbf{p}_m$$

advantages

- reduced dynamic range
- good numerical properties
- implementation with a systolic array







each iteration, K Givens rotation parameters have to be computed to transform the matrix-vector product into a zero vector to update the correlation matrix

updating process after the i -th Givens rotation

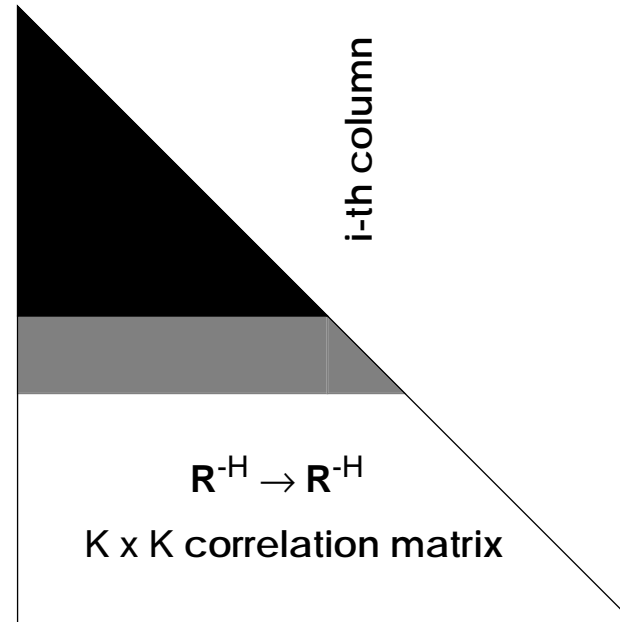
Kalman gain $0 \rightarrow -\delta \cdot k^H$



matrix-vector product

$-a \rightarrow 0$

i -th row



novel QRD based architecture for MRLS algorithms

express the recursive update equation for the modified correlation matrix as a matrix product

$$\begin{aligned} \mathbf{P}_{m+1, m} [l] &= (\mathbf{P}_{m, m-1} [l] - \zeta_m [l] \mathbf{k}_m^H [l]) + q_m \mathbf{1} = \mathbf{P}_m [l] + q_m [l] \mathbf{1} \\ &= \begin{bmatrix} \sqrt{q_m [l] \mathbf{1}} \\ \mathbf{R}_m^{-H} [l] \end{bmatrix}^H \begin{bmatrix} \sqrt{q_m [l] \mathbf{1}} \\ \mathbf{R}_m^{-H} [l] \end{bmatrix} \end{aligned}$$

\mathbf{R}^{-H} ... lower triangular matrix

new method:

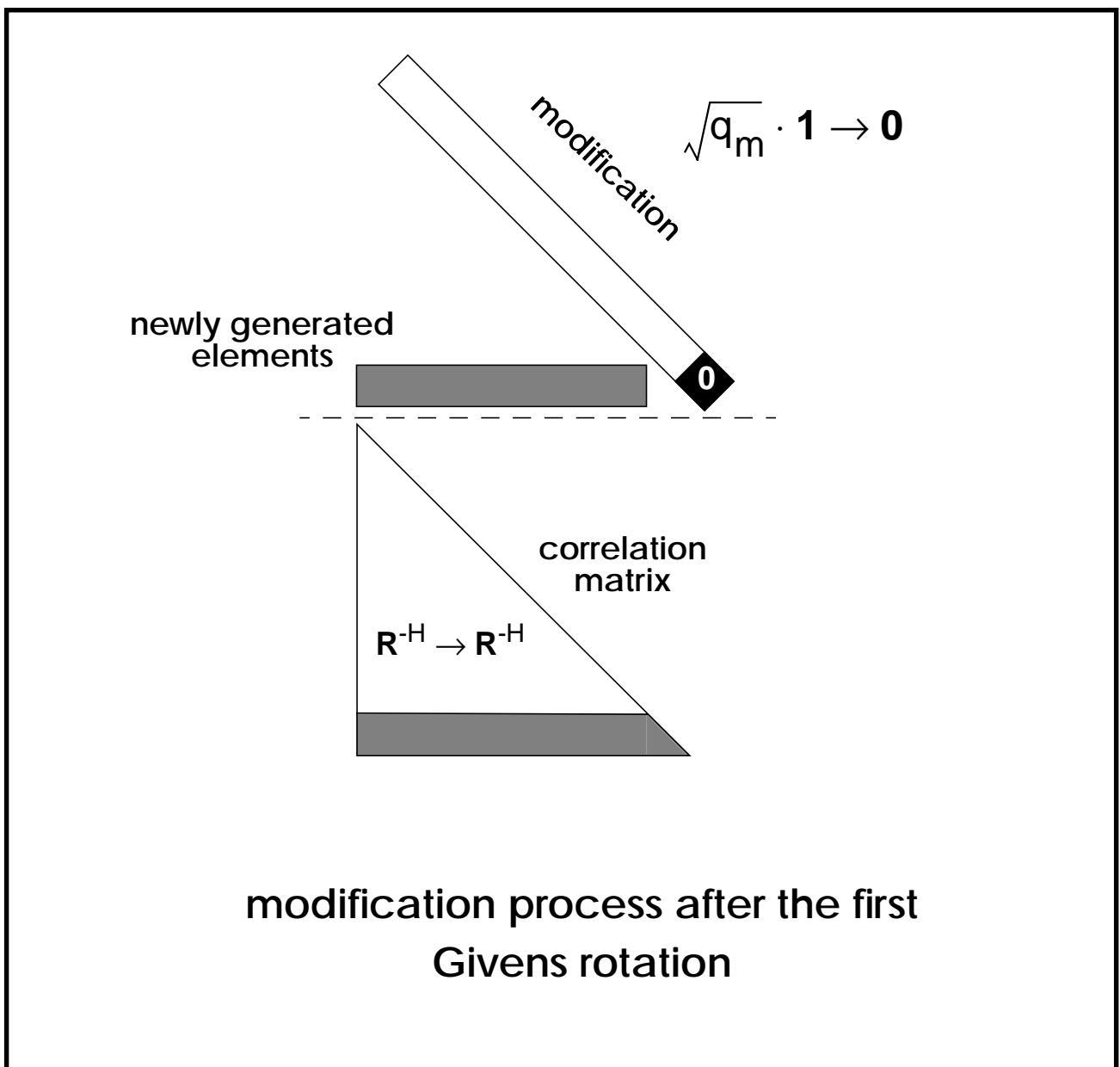
- ❑ Cholesky-factorization of the modified correlation matrix
- ❑ compute the Cholesky-factorization by means of Givens-rotations

$$\prod_{j=1}^J \Theta_{j, m} \cdot \begin{bmatrix} \sqrt{q_m [l] \mathbf{1}} \\ \mathbf{R}_m^{-H} [l] \end{bmatrix} = \begin{bmatrix} 0 \\ \mathbf{R}_{m+1, m}^{-H} [l] \end{bmatrix}$$

MRLS modified recursive least squares
QRD QR decomposition



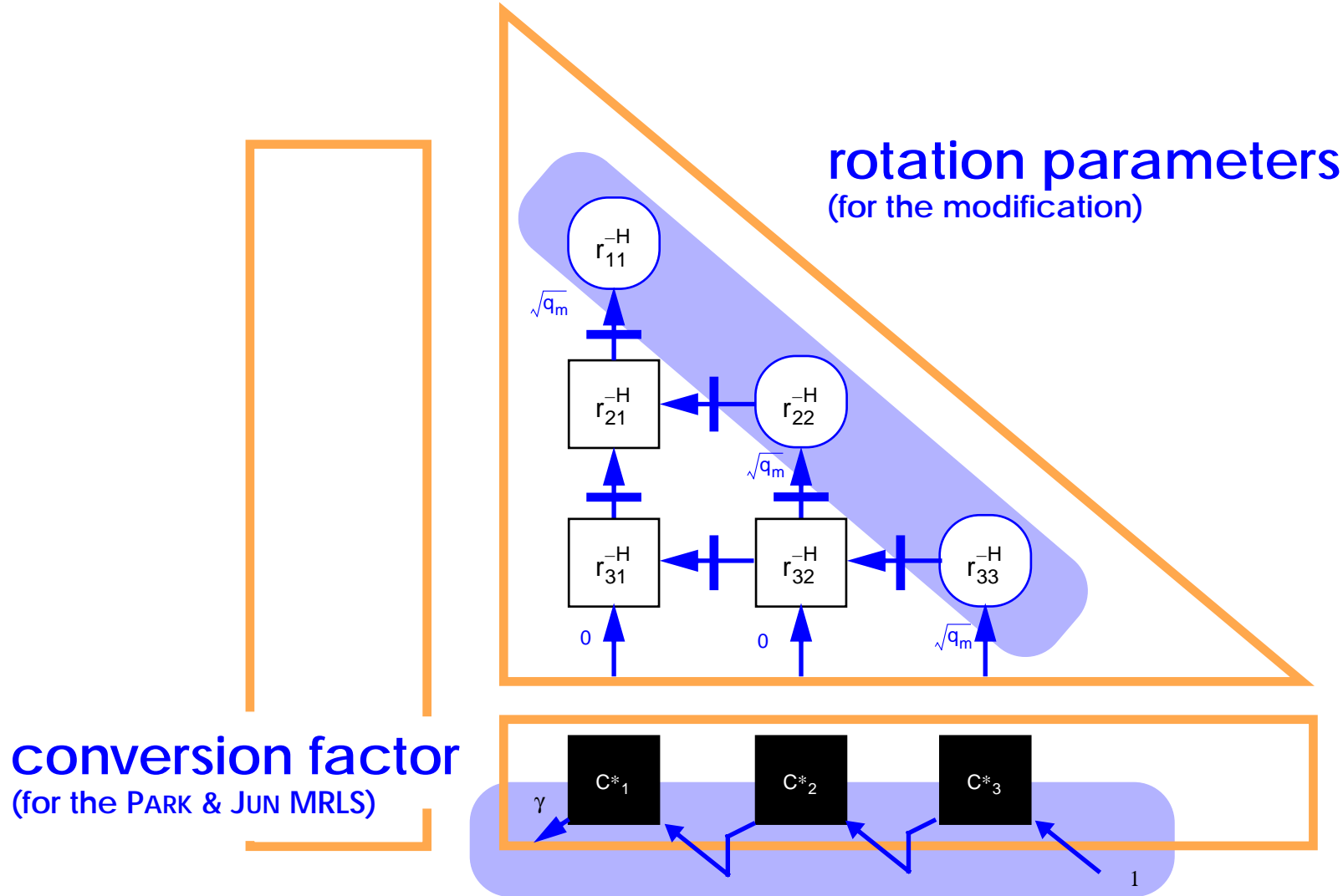
- for the modification, a diagonal matrix has to be rotated into the lower triangular correlation matrix
- this operation is similar to the correlation matrix update of the conventional QR RLS algorithm (GENTLEMAN & KUNG, 1981)



MRLS *modified recursive least squares*

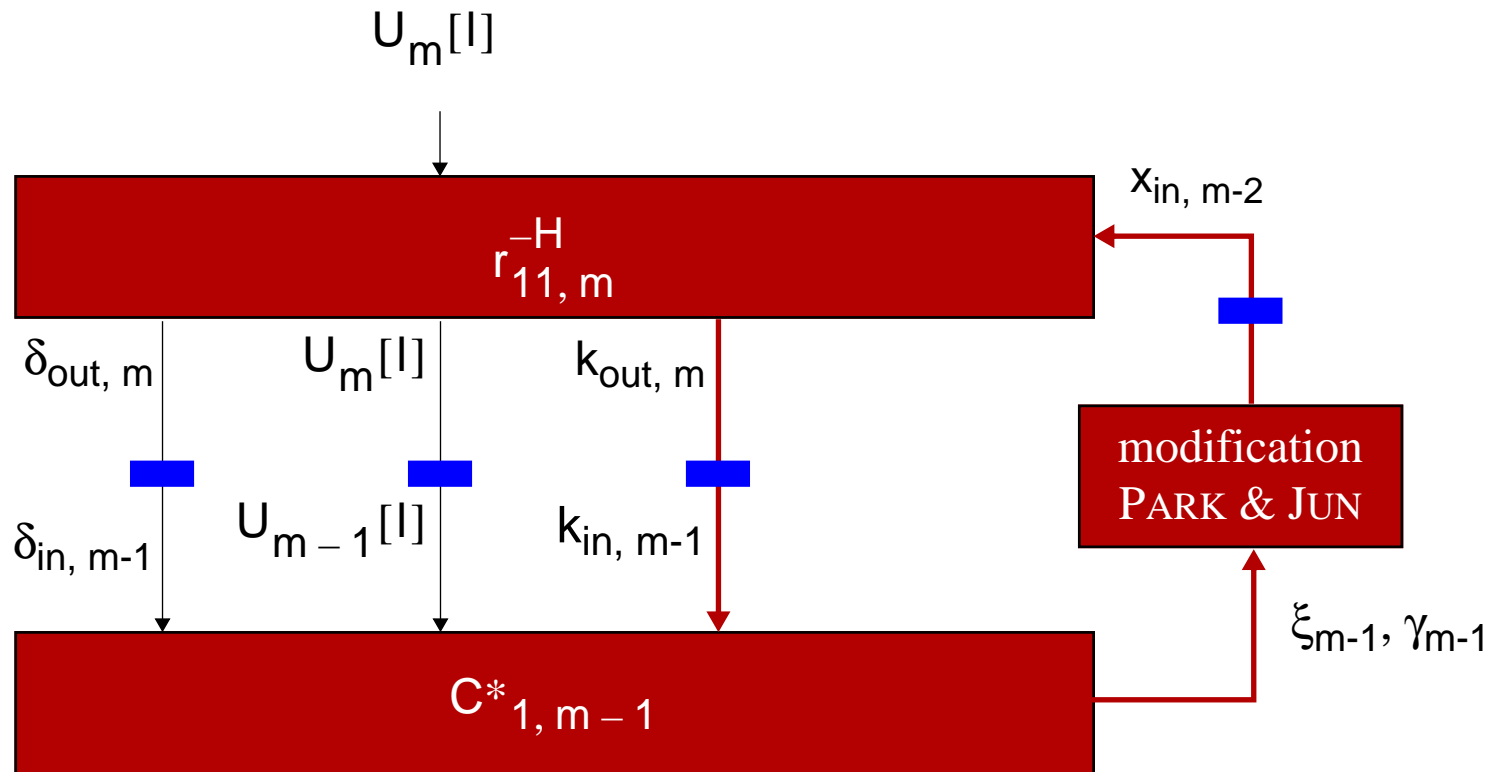
add the operations for the modification to MOONEN's array

MRLS
modified recursive least squares



The critical path contains the feedback loop for the modification.

- delay relaxation
- fully pipelined MRLS (PARK & JUN) for $K = 1$
- small degradation when the modification changes slowly with time



SER degradation due to finite precision

	$\langle W_U \rangle$	$\langle W_{max} \rangle$	SER (float)	target SER	SER (bit true)	
A	d.C.	12 Bit	$3,0 \cdot 10^{-3}$	$< 10^{-2}$	$3,4 \cdot 10^{-3}$	
	LMS	6 Bit	$< 2,0 \cdot 10^{-4}$	$< 10^{-2}$	$6,8 \cdot 10^{-3}$	
	QR-RLS	6 Bit	$< 2,0 \cdot 10^{-4}$	$< 10^{-2}$	$6,6 \cdot 10^{-3}$	
B	QR-MRLS	9 Bit	20 Bit	$5,3 \cdot 10^{-2}$	$5,3 \cdot 10^{-2}$	$4,8 \cdot 10^{-2}$

A ... phase noise + stationary fading channel,

B ... AWGN + time-variant fading channel

SER
d.C.
LMS
(M)RLS
FFT
AWGN

- differential coding imposes the highest precision requirements on the FFT output
- square values require high wordlengths

symbol error rate
differential coding
least mean squares
(modified) recursive least squares
fast Fourier transform
additive white gaussian noise



estimated complexity for FDE

	through-put [MHz]	clock [MHz]	NAND gate equivalents
d.C.	27,55	82,64	3 000
LMS	9,18	82,64	7 000
QR RLS r_{11}^{-H} cell	2,17	82,64	5 000
QR RLS C^*_1 cell	4,35	82,64	6 000
QR MRLS (delay relaxation) - r_{11}^{-H} cell	1,10	68,17	9 000
QR MRLS (delay relaxation) - C^*_1 cell	4,79	67,11	11 000
QR MRLS (no delay relaxation)	0,923	51,71	20 000

- differential coding has the lowest complexity
- high wordlengths reduce the throughput
- divisions and square-roots should be implemented as lookup-tables

d.C. differential coding
LMS least mean squares

(M)RLS (modified) recursive least squares
FDE frequency domain equalization

- **Phase noise introduces severe ICI.**
- **Linear adaptive FDE in combination with perfect guard intervals provides more flexibility than coherent channel estimation and is less sensitive with respect to ICI than differential coding.**
- **A novel QRD based architecture for the MRLS algorithm has been developed.**
- **Neither the RLS, LMS or MRLS algorithms, nor differential coding, can mitigate the requirements on the purity of oscillators in OFDM-systems significantly.**
- **From a complexity and performance point of view, QR RLS FDE or differential coding are the methods of choice for FRA subject to moderate oscillator phase noise.**
- **Differential coding has the lowest complexity but imposes the highest precision requirements on the FFT output.**
- **Adaptive FDEs are applicable to OFDMA-systems and non-orthogonal filter banks with better spectral properties than DFT filter banks, as well.**

