



Escola Tècnica Superior d'Enginyeria de  
Telecomunicació de Barcelona



UNIVERSITAT POLITÈCNICA  
DE CATALUNYA  
BARCELONATECH

The background features a light blue and white color scheme with a pattern of binary code (0s and 1s) scattered across the surface. Two globes of the Earth are visible, one in the center-left and one in the center-right, both rendered in a dark blue silhouette. The title "FIBER-OPTIC COMMUNICATIONS" is prominently displayed in the center-right in a large, bold, red font.

# FIBER-OPTIC COMMUNICATIONS

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[www.tsc.upc.edu/gco](http://www.tsc.upc.edu/gco)

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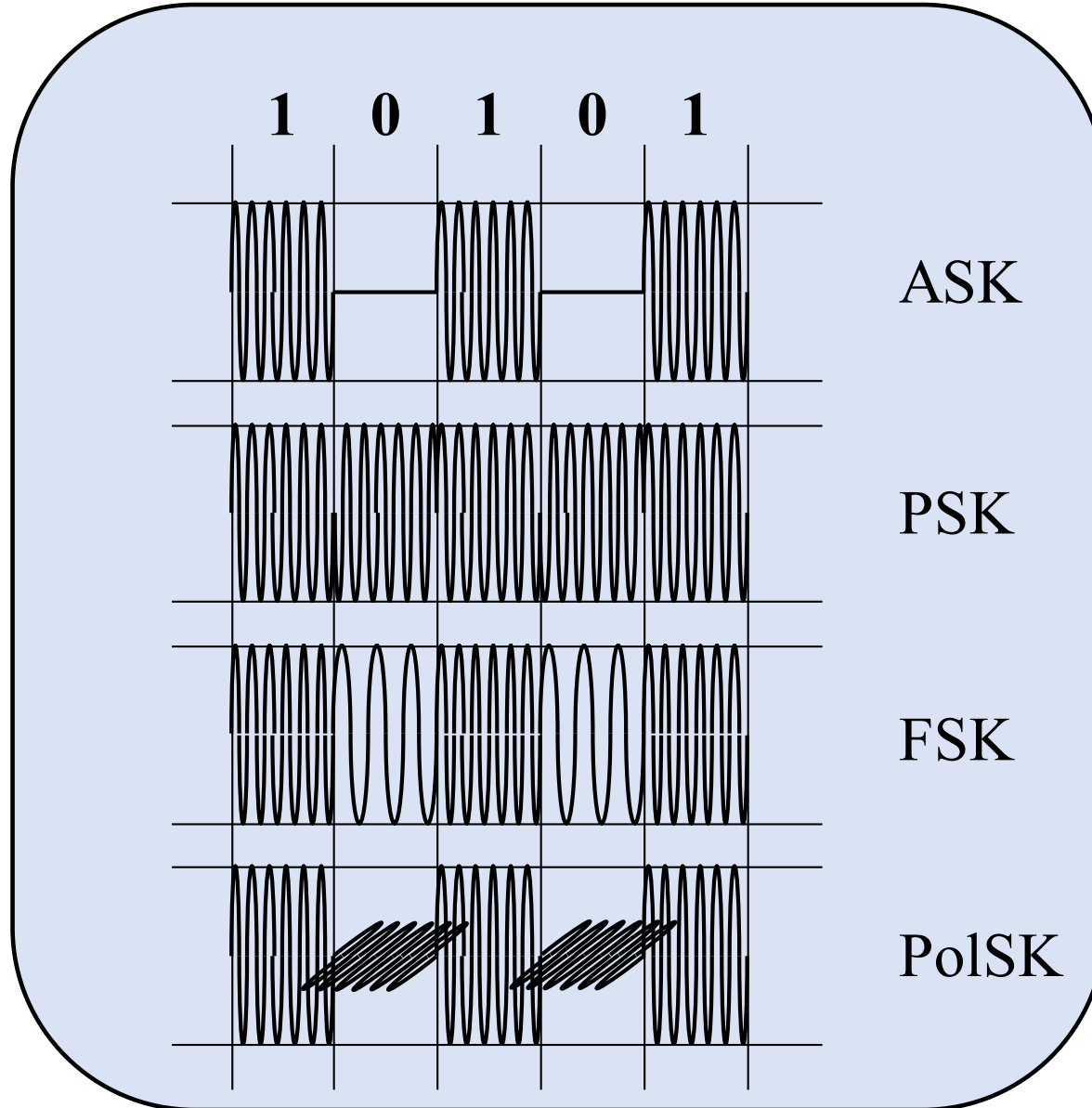
# 7. COHERENT DETECTION

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- 
- The background features a grayscale world map centered on Europe, overlaid on a pattern of binary code (0s and 1s) that appears to be floating or falling around the map.
- ❑ **DIGITAL COHERENT DETECTION**
    - ❑ **CHROMATIC DISPERSION COMPENSATION**
    - ❑ **FREQUENCY & PHASE ESTIMATION**
    - ❑ **POLARIZATION TRACKING AND PMD COMPENSATION**
  - ❑ **RECENT EXPERIMENTS & COMMERCIAL EQUIPMENT**

**MODULATION OF LIGHT PROPERTIES**



amplitude

phase

frequency

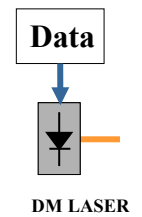
polarization

# Modulation/ Detection Schemes

## Modulation

Direct

Intensity  
Frequency  
Phase



External

EAM

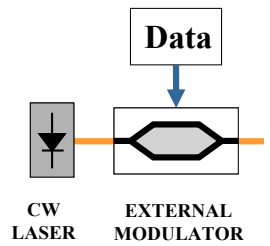
Intensity

ERM

Phase  
Polarization

Mach-Zehnder

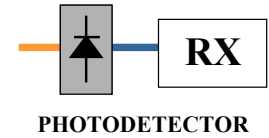
Intensity  
Phase



## Detection

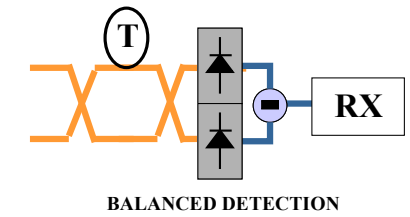
Direct

Intensity



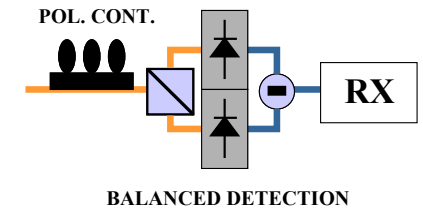
Differential

Phase  
Frequency  
Polarization



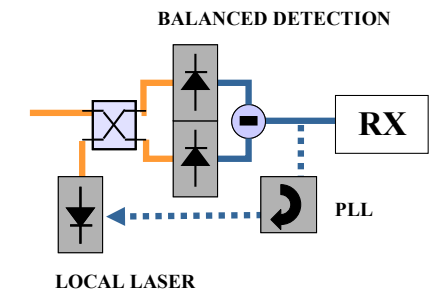
Polarization

Polarization



Synchronous

Intensity  
Phase  
Frequency  
Polarization



# COHERENT DETECTION CONCEPT

$$e_s(t) = A_s(t) e^{j\omega_s t} = |A_s(t)| e^{j(\omega_s t + \theta_s(t))}$$

$$e_L(t) = A_L e^{j\omega_L t} = |A_L| e^{j\left(\omega_L t + \theta_L(t) - \frac{\pi}{2}\right)}$$

Output Signal

information

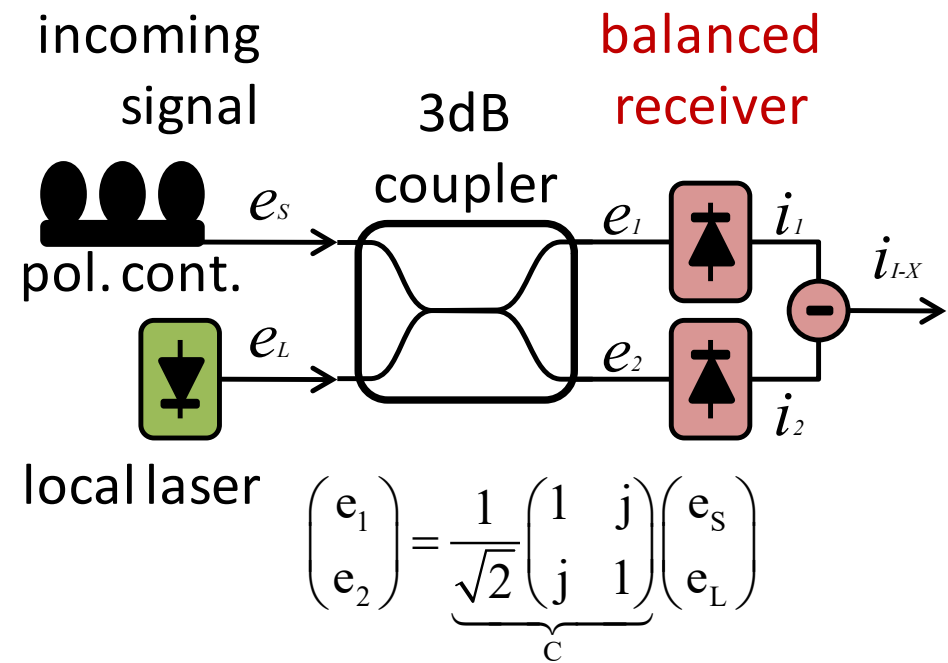
$$i_{I-X} = 2R |A_s(t)| |A_L| \cos(\omega_{FI} t + \theta(t))$$

$$\theta(t) \equiv \theta_s(t) - \theta_L(t)$$

$$\omega_{FI} \equiv \omega_s - \omega_L$$

intermediate frequency

heterodyne detection  $\omega_{FI} \neq 0$   
 homodyne detection  $\omega_{FI} = 0$

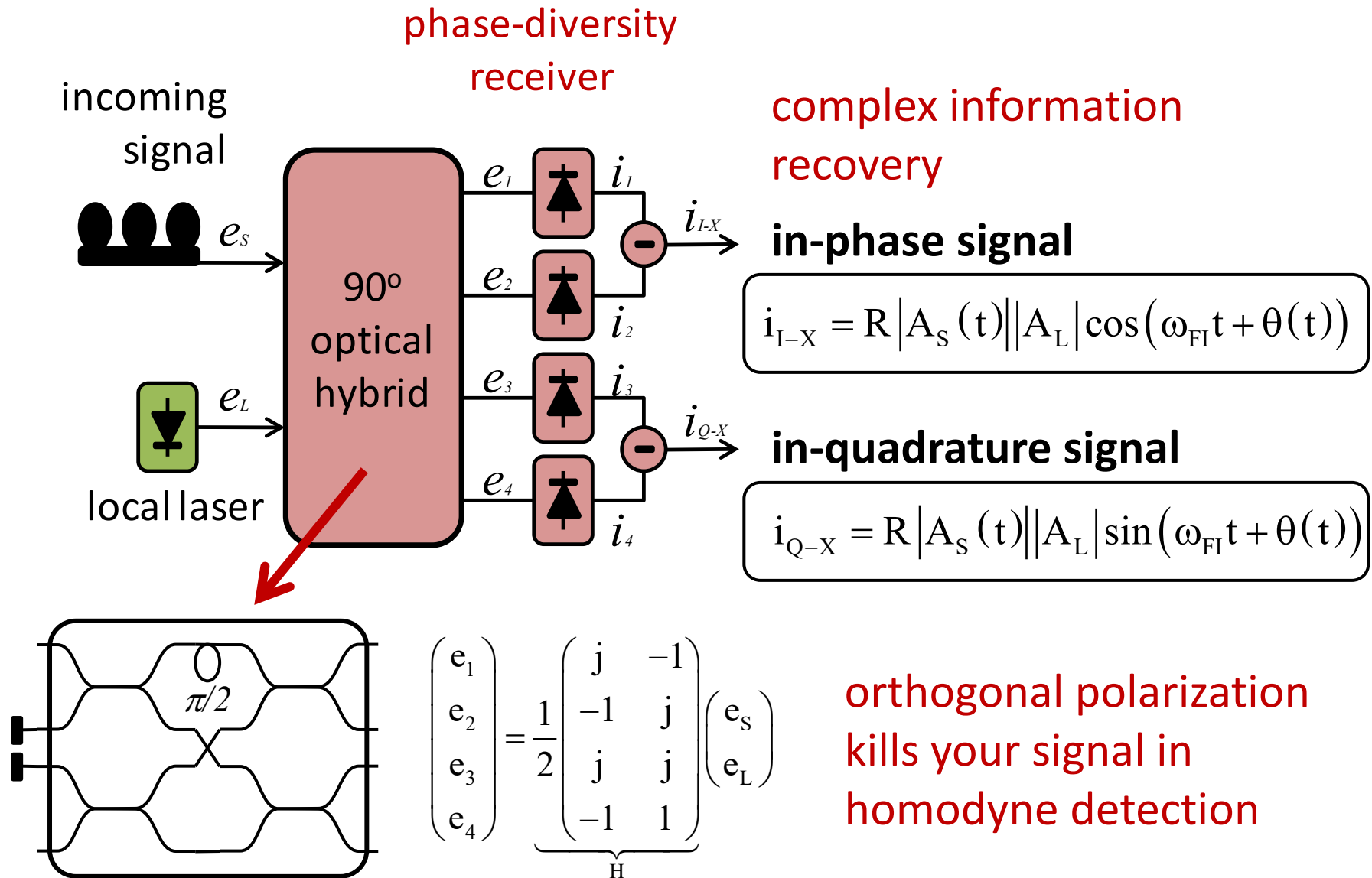


a phase error of  $\pi/2$  kills your signal in homodyne detection





# Phase-Diversity Coherent Receiver



# Mathematical Development

$$e_s(t) = A_s(t) e^{j\omega_s t} = |A_s(t)| e^{j(\omega_s t + \theta_s(t))}$$

$$e_L(t) = A_L e^{j\omega_L t} = |A_L| e^{j(\omega_L t + \theta_L(t) - \frac{\pi}{2})}$$

$$\begin{pmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \end{pmatrix} = \frac{1}{2} \underbrace{\begin{pmatrix} j & -1 \\ -1 & j \\ j & j \\ -1 & 1 \end{pmatrix}}_H \begin{pmatrix} e_s \\ e_L \end{pmatrix}$$

$$e_1 = \frac{1}{2}(j e_s - e_L) = \frac{1}{2} \left( |A_s(t)| e^{j(\omega_s t + \frac{\pi}{2})} - |A_L| e^{j(\omega_L t - \frac{\pi}{2})} \right)$$

$$e_2 = \frac{1}{2}(-e_s + j e_L) = \frac{1}{2} \left( -|A_s(t)| e^{j\omega_s t} + |A_L| e^{j\omega_L t} \right)$$

$$e_3 = \frac{j}{2}(e_s + e_L) = \frac{j}{2} \left( |A_s(t)| e^{j\omega_s t} + |A_L| e^{j(\omega_L t - \frac{\pi}{2})} \right)$$

$$e_4 = \frac{1}{2}(-e_s + e_L) = \frac{1}{2} \left( -|A_s(t)| e^{j\omega_s t} + |A_L| e^{j(\omega_L t - \frac{\pi}{2})} \right)$$

$$i_1 = R |e_3|^2 = R \left[ \frac{|A_s(t)|^2}{4} + \frac{|A_L|^2}{4} - \frac{|A_s(t)||A_L|}{2} \underbrace{\text{Re} \left\{ e^{j(\omega_s t + \theta_s(t) + \frac{\pi}{2})} e^{-j(\omega_L t + \theta_L(t) - \frac{\pi}{2})} \right\}}_{\substack{\cos((\omega_s - \omega_L)t + \theta_s(t) - \theta_L(t) + \pi) \\ -\cos(\omega_{FI}t + \theta(t))}} \right]$$

$$i_2 = R |e_4|^2 = R \left[ \frac{|A_s(t)|^2}{4} + \frac{|A_L|^2}{4} - \frac{|A_s(t)||A_L|}{2} \underbrace{\text{Re} \left\{ e^{j(\omega_s t + \theta_s(t))} e^{-j(\omega_L t + \theta_L(t))} \right\}}_{\substack{\cos((\omega_s - \omega_L)t + \theta_s(t) - \theta_L(t)) \\ \cos(\omega_{FI}t + \theta(t))}} \right]$$

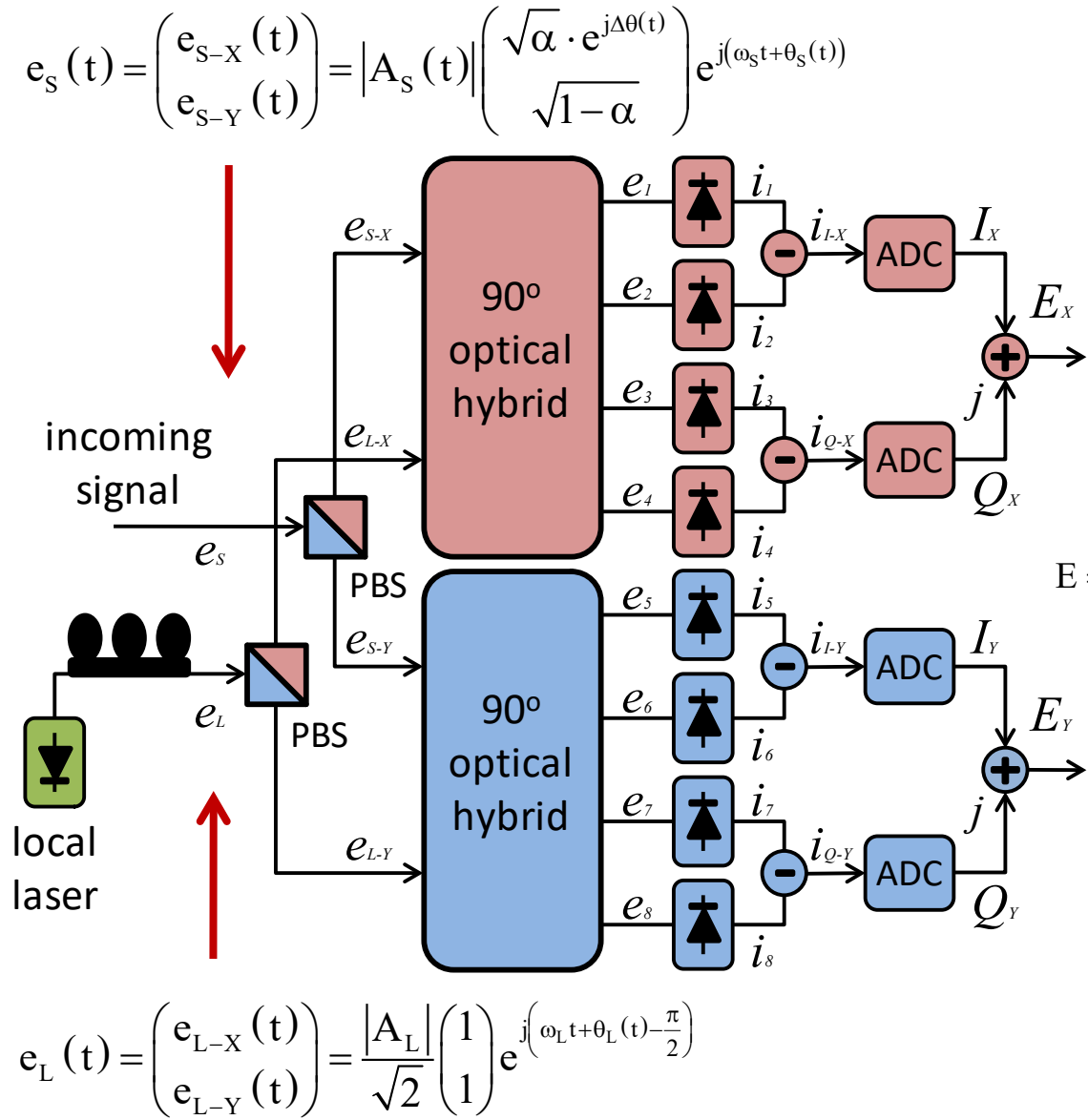
$$i_3 = R |e_3|^2 = R \left[ \frac{|A_s(t)|^2}{4} + \frac{|A_L|^2}{4} + \frac{|A_s(t)||A_L|}{2} \underbrace{\text{Re} \left\{ e^{j(\omega_s t + \theta_s(t))} e^{-j(\omega_L t + \theta_L(t) - \frac{\pi}{2})} \right\}}_{\substack{\cos((\omega_s - \omega_L)t + \theta_s(t) - \theta_L(t) + \frac{\pi}{2}) \\ \sin(\omega_{FI}t + \theta(t))}} \right]$$

$$i_4 = R |e_4|^2 = R \left[ \frac{|A_s(t)|^2}{4} + \frac{|A_L|^2}{4} - \frac{|A_s(t)||A_L|}{2} \underbrace{\text{Re} \left\{ e^{j(\omega_s t + \theta_s(t))} e^{-j(\omega_L t + \theta_L(t) - \frac{\pi}{2})} \right\}}_{\substack{\cos((\omega_s - \omega_L)t + \theta_s(t) - \theta_L(t) + \frac{\pi}{2}) \\ \sin(\omega_{FI}t + \theta(t))}} \right]$$

$$i_{I-X} = i_1 - i_2 = R |A_s(t)||A_L| \cos(\omega_{FI}t + \theta(t))$$

$$i_{Q-X} = i_3 - i_4 = R |A_s(t)||A_L| \sin(\omega_{FI}t + \theta(t))$$

# Phase & Polarization-Diversity



full information recovery

$$E_X = R \sqrt{\frac{\alpha}{2}} |A_s(t)| |A_L| e^{j(\omega_{FI} t + \theta(t) + \Delta\theta(t))}$$

X-polarization signal

$$E = \frac{R}{\sqrt{2}} |A_s(t)| |A_L| \begin{pmatrix} \sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \\ \sqrt{1-\alpha} \end{pmatrix} e^{j(\omega_{FI} t + \theta_s(t) - \theta_L(t))}$$

Y-polarization signal

$$E_Y = R \sqrt{\frac{1-\alpha}{2}} |A_s(t)| |A_L| e^{j(\omega_{FI} t + \theta(t))}$$

# Mathematical Development (I)

$$\mathbf{e}_s(t) = \begin{pmatrix} e_{s-X}(t) \\ e_{s-Y}(t) \end{pmatrix} = \begin{pmatrix} A_{s-X}(t) \\ A_{s-Y}(t) \end{pmatrix} e^{j\omega_s t} = |A_s(t)| \begin{pmatrix} \sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \\ \sqrt{1-\alpha} \end{pmatrix} e^{j(\omega_s t + \theta_s(t))}$$

$$\mathbf{e}_L(t) = \begin{pmatrix} e_{L-X}(t) \\ e_{L-Y}(t) \end{pmatrix} = \begin{pmatrix} A_{L-X}(t) \\ A_{L-Y}(t) \end{pmatrix} e^{j\omega_L t} = \frac{|A_L|}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} e^{j(\omega_L t + \theta_L(t) - \frac{\pi}{2})}$$

$$e_1 = \frac{1}{2}(j e_{s-X} - e_{L-X}) = \frac{1}{2} \left( \sqrt{\alpha} |A_s(t)| e^{j(\omega_s t + \theta_s(t) + \Delta\theta(t) + \frac{\pi}{2})} - \frac{|A_L|}{\sqrt{2}} e^{j(\omega_L t + \theta_L(t) - \frac{\pi}{2})} \right)$$

$$\begin{pmatrix} e_1 & e_5 \\ e_2 & e_6 \\ e_3 & e_7 \\ e_4 & e_8 \end{pmatrix} = \frac{1}{2} \underbrace{\begin{pmatrix} j & -1 \\ -1 & j \\ j & j \\ -1 & 1 \end{pmatrix}}_H \begin{pmatrix} e_s^T \\ e_L^T \end{pmatrix}$$

$$e_2 = \frac{1}{2}(-e_{s-X} + j e_{L-X}) = \frac{1}{2} \left( -\sqrt{\alpha} |A_s(t)| e^{j(\omega_s t + \theta_s(t) + \Delta\theta(t))} + \frac{|A_L|}{\sqrt{2}} e^{j(\omega_L t + \theta_L(t))} \right)$$

$$e_3 = \frac{j}{2}(e_{s-X} + e_{L-X}) = \frac{j}{2} \left( \sqrt{\alpha} |A_s(t)| e^{j(\omega_s t + \theta_s(t) + \Delta\theta(t))} + \frac{|A_L|}{\sqrt{2}} e^{j(\omega_L t + \theta_L(t) - \frac{\pi}{2})} \right)$$

$$e_4 = \frac{1}{2}(-e_{s-X} + e_{L-X}) = \frac{1}{2} \left( -\sqrt{\alpha} |A_s(t)| e^{j(\omega_s t + \theta_s(t) + \Delta\theta(t))} + \frac{|A_L|}{\sqrt{2}} e^{j(\omega_L t + \theta_L(t) - \frac{\pi}{2})} \right)$$

$$e_5 = \frac{1}{2}(j e_{s-Y} - e_{L-Y}) = \frac{1}{2} \left( \sqrt{1-\alpha} |A_s(t)| e^{j(\omega_s t + \theta_s(t) + \frac{\pi}{2})} - \frac{|A_L|}{\sqrt{2}} e^{j(\omega_L t + \theta_L(t) - \frac{\pi}{2})} \right)$$

$$e_6 = \frac{1}{2}(-e_{s-Y} + j e_{L-Y}) = \frac{1}{2} \left( -\sqrt{1-\alpha} |A_s(t)| e^{j(\omega_s t + \theta_s(t))} + \frac{|A_L|}{\sqrt{2}} e^{j(\omega_L t + \theta_L(t))} \right)$$





# Mathematical Development (III)

$$i_{I-X} = i_1 - i_2 = R \sqrt{\frac{\alpha}{2}} |A_S(t)| |A_L| \cos(\omega_{FI} t + \theta(t) + \Delta\theta(t))$$

$$i_{Q-X} = i_3 - i_4 = R \sqrt{\frac{\alpha}{2}} |A_S(t)| |A_L| \sin(\omega_{FI} t + \theta(t) + \Delta\theta(t))$$

$$i_{I-Y} = i_5 - i_6 = R \sqrt{\frac{1-\alpha}{2}} |A_S(t)| |A_L| \cos(\omega_{FI} t + \theta(t))$$

$$i_{Q-Y} = i_7 - i_8 = R \sqrt{\frac{1-\alpha}{2}} |A_S(t)| |A_L| \sin(\omega_{FI} t + \theta(t))$$

$$i_X = i_{I-X} + j \cdot i_{Q-X} = R \sqrt{\frac{\alpha}{2}} |A_S(t)| |A_L| e^{j(\omega_{FI} t + \theta(t) + \Delta\theta(t))}$$

$$i_Y = i_{I-Y} + j \cdot i_{Q-Y} = R \sqrt{\frac{1-\alpha}{2}} |A_S(t)| |A_L| e^{j(\omega_{FI} t + \theta(t))}$$

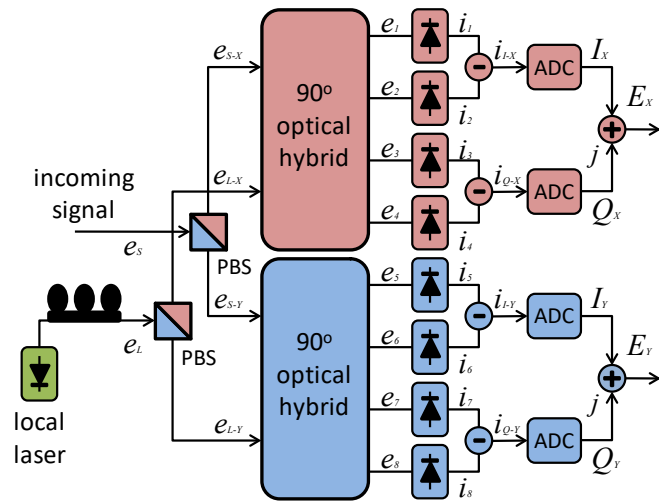
$$E = \frac{R}{\sqrt{2}} |A_S(t)| |A_L| \begin{pmatrix} \sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \\ \sqrt{1-\alpha} \end{pmatrix} e^{j(\omega_{FI} t + \overbrace{\theta_S(t) - \theta_L(t)}^{\theta(t)})}$$

# Polarization Division Multiplexing (PDM)

Channel  $\parallel$   
 $e_s(t) = e_s^{\parallel}(t) + e_s^{\perp}(t) \rightarrow$   
Channel  $\perp$

$$\begin{cases} e_s^{\parallel}(t) = |A_S^{\parallel}(t)| \begin{pmatrix} \sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \\ \sqrt{1-\alpha} \end{pmatrix} e^{j(\omega_s t + \theta_S^{\parallel}(t))} \\ e_s^{\perp}(t) = |A_S^{\perp}(t)| \begin{pmatrix} \sqrt{1-\alpha} \\ -\sqrt{\alpha} \cdot e^{-j\Delta\theta(t)} \end{pmatrix} e^{j(\omega_s t + \theta_S^{\perp}(t))} \end{cases}$$

orthogonal components



$$E = E_{\parallel} + E_{\perp}$$

$$\begin{cases} E_{\parallel} = \frac{R}{\sqrt{2}} |A_S^{\parallel}(t)| |A_L| \begin{pmatrix} \sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \\ \sqrt{1-\alpha} \end{pmatrix} e^{j(\omega_{FI} t + \overbrace{\theta_S^{\parallel}(t) - \theta_L(t)}^{\theta_{\parallel}(t)})} \\ E_{\perp} = \frac{R}{\sqrt{2}} |A_S^{\perp}(t)| |A_L| \begin{pmatrix} \sqrt{1-\alpha} \\ -\sqrt{\alpha} \cdot e^{-j\Delta\theta(t)} \end{pmatrix} e^{j(\omega_{FI} t + \overbrace{\theta_S^{\perp}(t) - \theta_L(t)}^{\theta_{\perp}(t)})} \end{cases}$$

mixed information

$$e_L(t) = \begin{pmatrix} e_{L-X}(t) \\ e_{L-Y}(t) \end{pmatrix} = \frac{|A_L|}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} e^{j(\omega_L t + \theta_L(t) - \frac{\pi}{2})}$$

# Polarization Division Multiplexing (PDM)

$$E_{\parallel} = \frac{R}{\sqrt{2}} |A_S^{\parallel}(t)| |A_L| \begin{pmatrix} \sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \\ \sqrt{1-\alpha} \end{pmatrix} e^{j(\omega_{FI}t + \theta_S^{\parallel}(t) - \theta_L(t))}$$

$$E_{\perp} = \frac{R}{\sqrt{2}} |A_S^{\perp}(t)| |A_L| \begin{pmatrix} \sqrt{1-\alpha} \\ -\sqrt{\alpha} \cdot e^{-j\Delta\theta(t)} \end{pmatrix} e^{j(\omega_{FI}t + \theta_S^{\perp}(t) - \theta_L(t))}$$

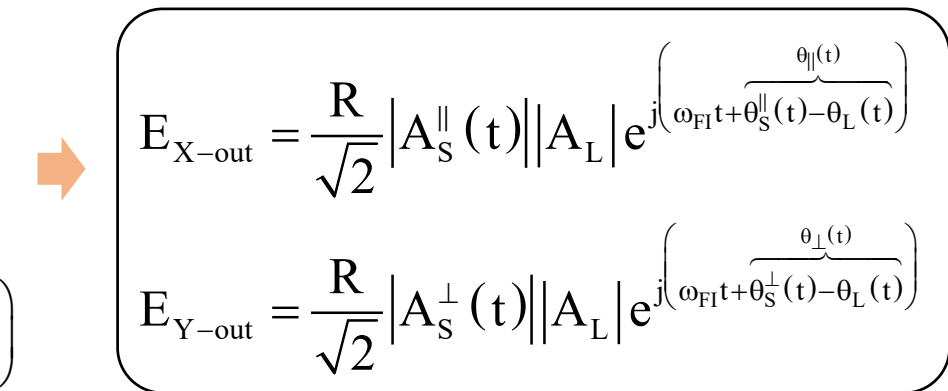
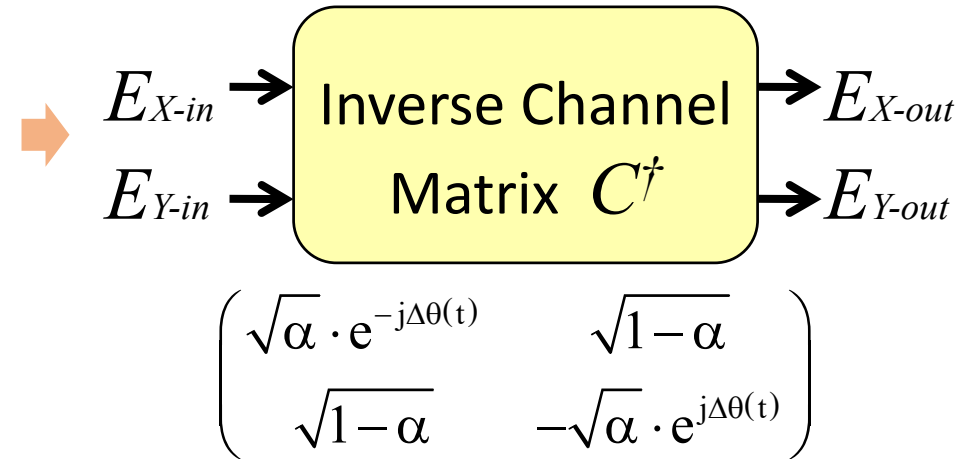
$$C^{\dagger} (E_{\parallel} + E_{\perp}) \rightarrow A_{\parallel} \begin{pmatrix} 1 \\ 0 \end{pmatrix} + A_{\perp} \begin{pmatrix} 0 \\ -1 \end{pmatrix}$$

$$\begin{pmatrix} \sqrt{\alpha} \cdot e^{-j\Delta\theta(t)} & \sqrt{1-\alpha} \\ \sqrt{1-\alpha} & -\sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \end{pmatrix} \begin{pmatrix} \sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \\ \sqrt{1-\alpha} \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} \sqrt{\alpha} \cdot e^{-j\Delta\theta(t)} & \sqrt{1-\alpha} \\ \sqrt{1-\alpha} & -\sqrt{\alpha} \cdot e^{j\Delta\theta(t)} \end{pmatrix} \begin{pmatrix} \sqrt{1-\alpha} \\ -\sqrt{\alpha} \cdot e^{-j\Delta\theta(t)} \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

PDL: Polarization-Dependent Loss

unitary channel  
(no PDL\*)



recovered information

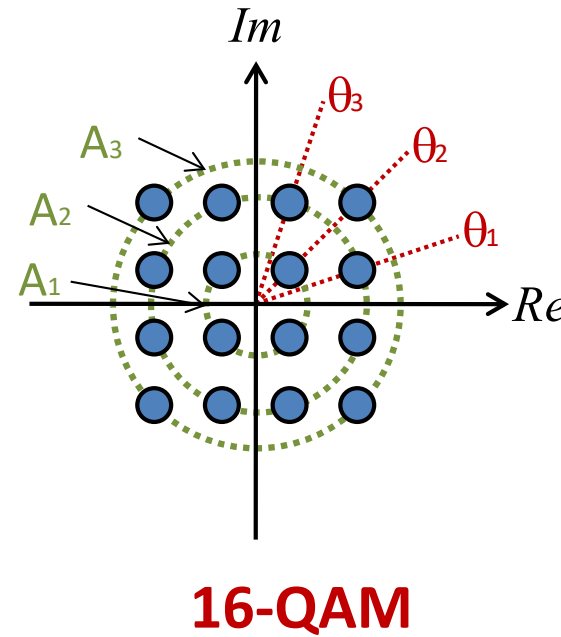
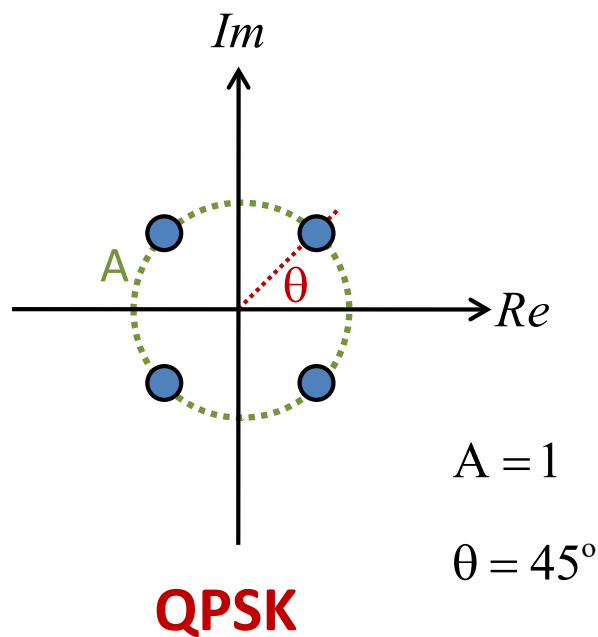
# Polarization Division Multiplexing (PDM)

$$E_{X-out} = \frac{R}{\sqrt{2}} |A_S^{\parallel}(t)| |A_L| e^{j(\omega_{FI}t + \overbrace{\theta_S^{\parallel}(t) - \theta_L(t)}^{\theta_{\parallel}(t)})}$$

$$E_{Y-out} = \frac{R}{\sqrt{2}} |A_S^{\perp}(t)| |A_L| e^{j(\omega_{FI}t + \overbrace{\theta_S^{\perp}(t) - \theta_L(t)}^{\theta_{\perp}(t)})}$$

## Modulation Examples

Normalized Power:  $P = A^2$



$$A_1 = \frac{A}{\sqrt{5}} \approx 0.45 \cdot A$$

$$A_2 = A$$

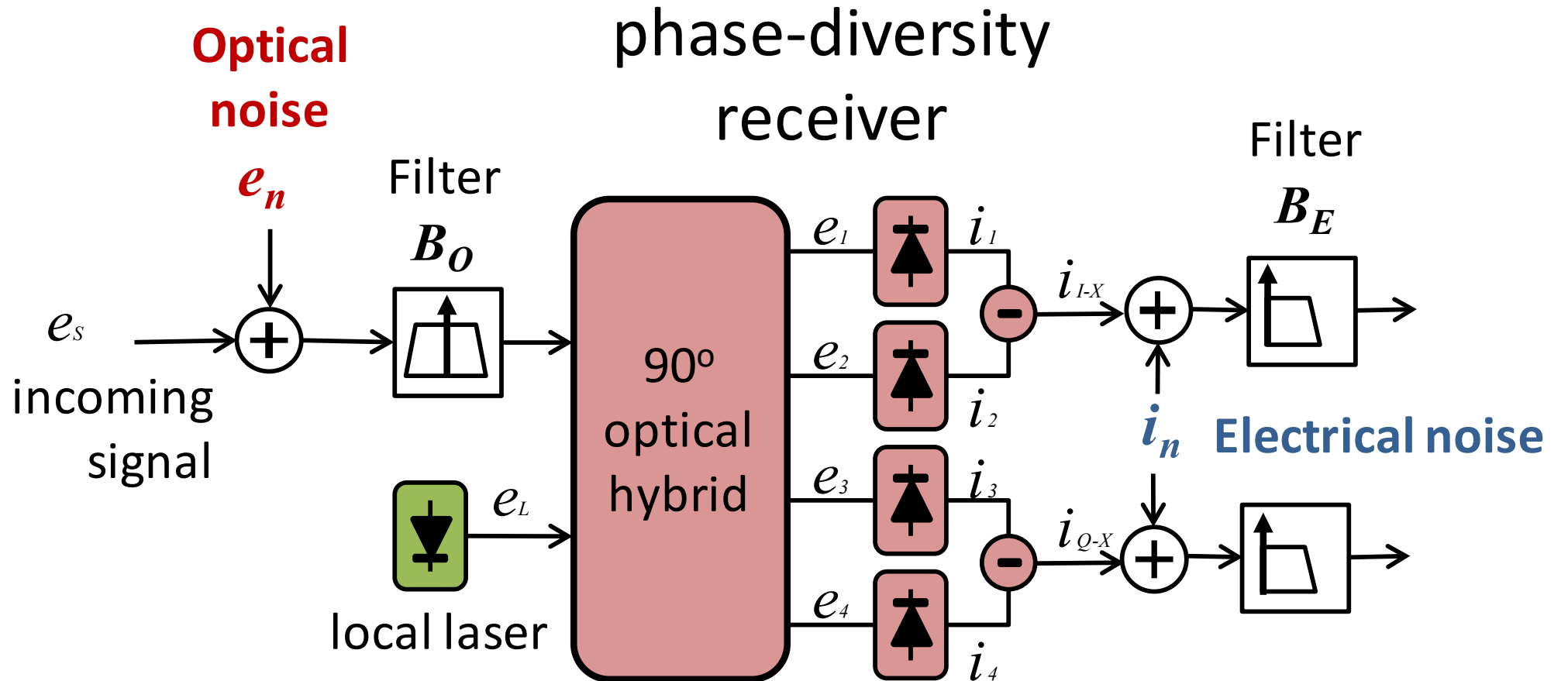
$$A_3 = \frac{3A}{\sqrt{5}} \approx 1.34 \cdot A$$

$$\theta_1 \approx 18^\circ$$

$$\theta_2 = 45^\circ$$

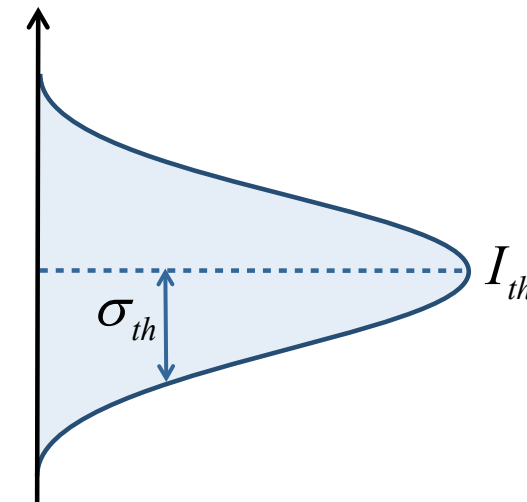
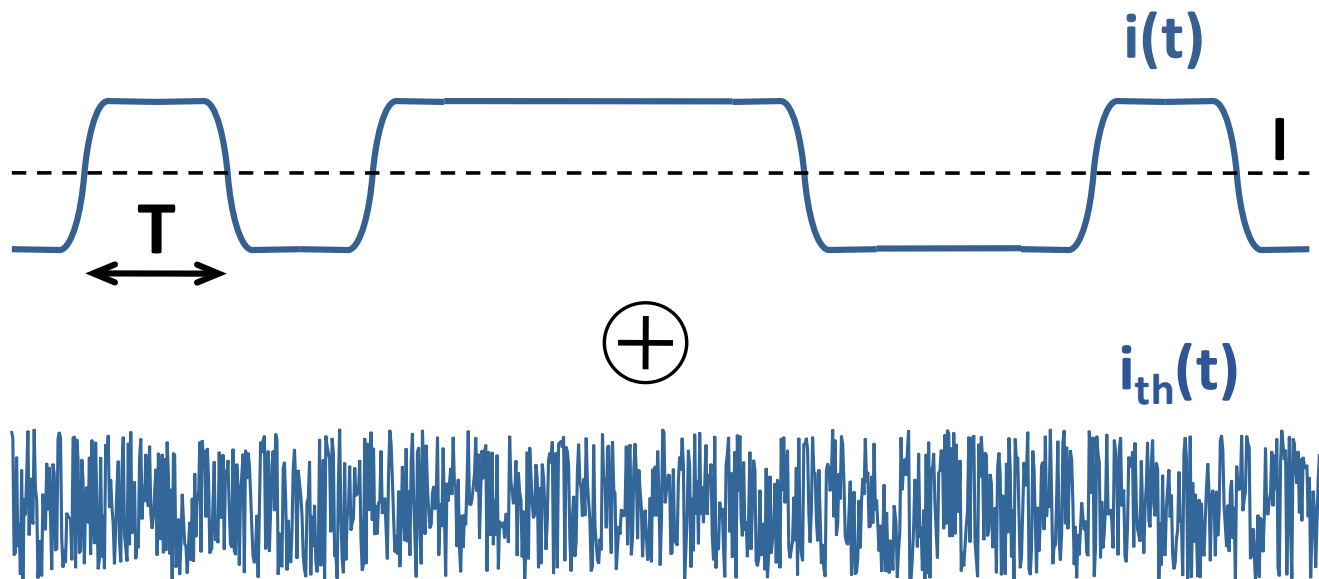
$$\theta_3 \approx 72^\circ$$

# NOISE IN COHERENT DETECTION





# Electrical Noise: Thermal



## AWGN Process

$$I_{th} = 0 \quad [A]$$

$$\sigma_{th}^2 = 4 \underbrace{\frac{K_B T}{R_L}}_{S_{th}} B_e \quad [A^2]$$

(Typical Value)

$2 \cdot 10^{-22}$

←  $S_{th}$ : Noise Spectral Density [ $A^2/Hz$ ]

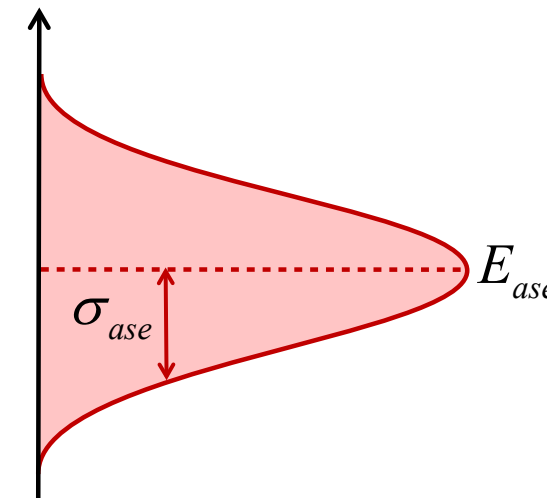
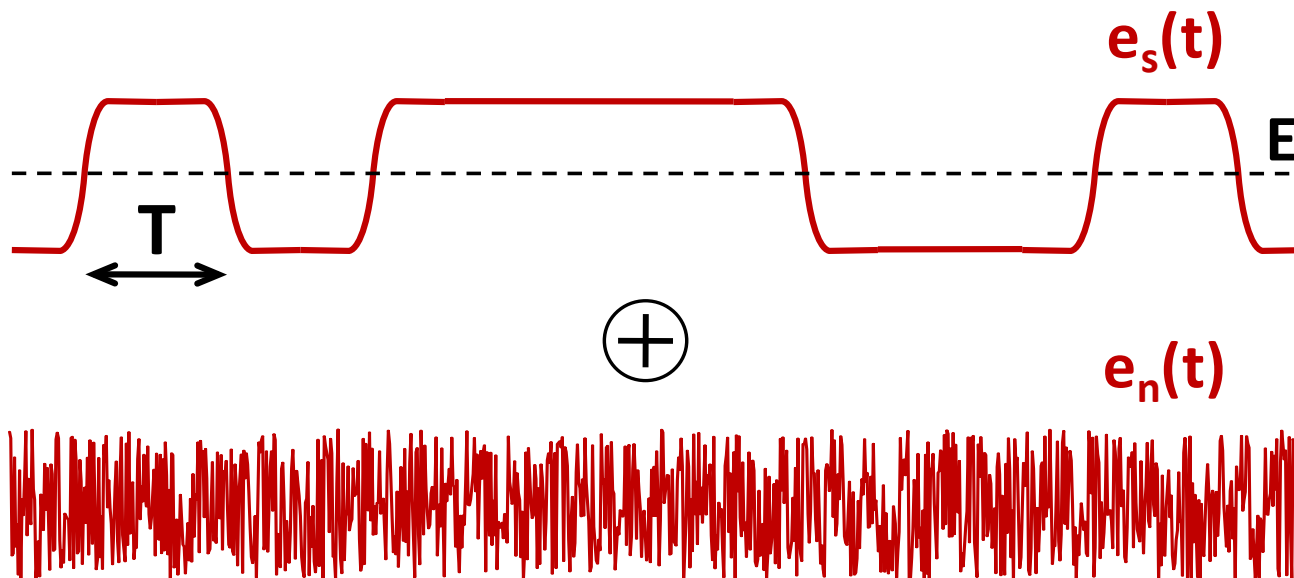
$K_B$ : Boltzmann's Constant [ $J/K$ ]

$T$ : Temperature [ $K$ ]

$R_L$ : Circuit Load [ $\Omega$ ]

$B_e$ : Noise Equivalent Bandwidth [ $Hz$ ]

# Optical Noise: ASE



## AWGN Process

$$E_{ase} = 0$$

$$\underbrace{\sigma_{ase}^2}_{P_{ase}} \approx S_{ase}(f_c) B_o$$

$$[W^{1/2}]$$

$$[W]$$

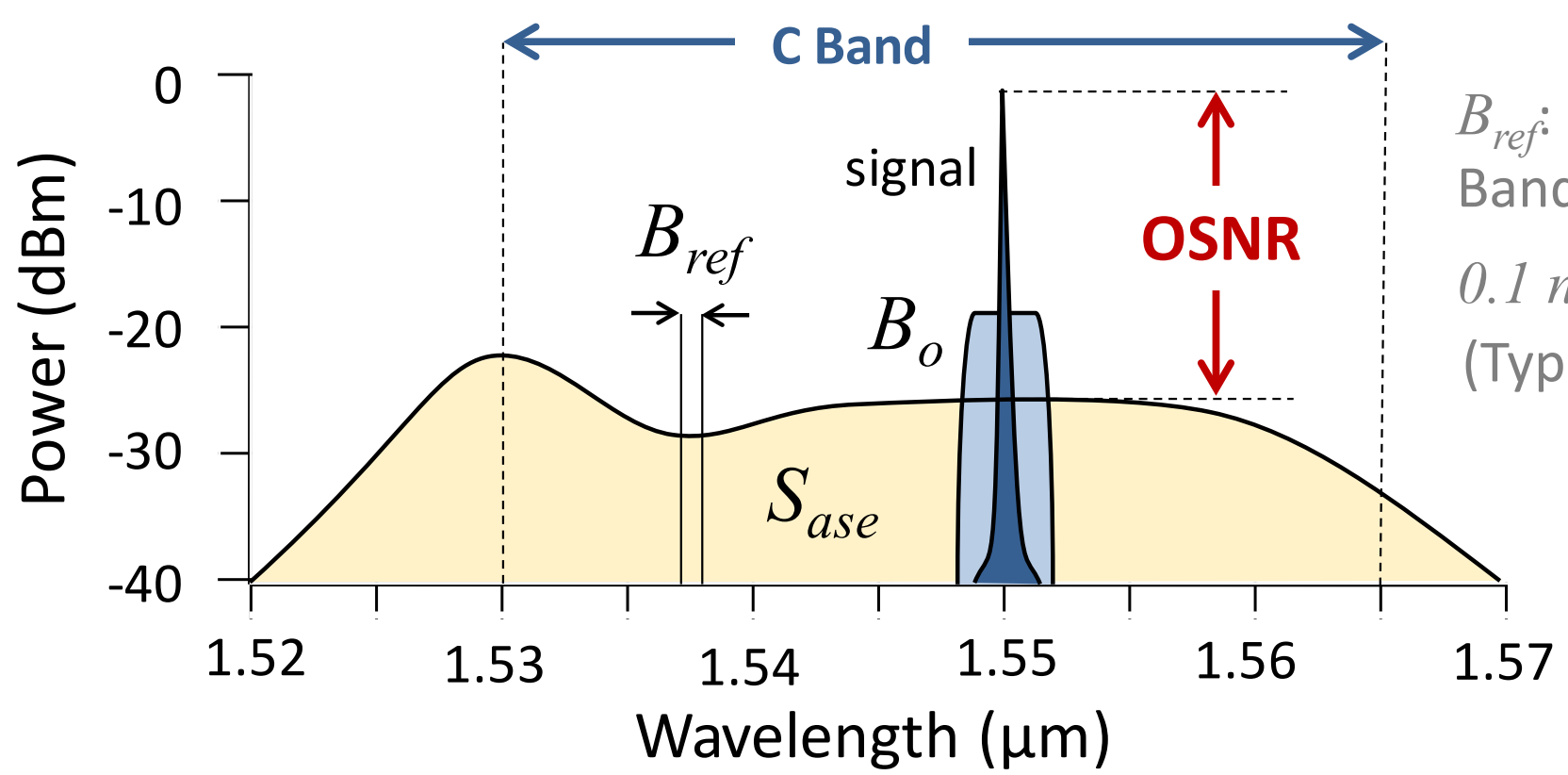
ASE: Amplified Spontaneous Emission

$S_{ase}$ : ASE Spectral Density  $[W/Hz]$

$B_o$ : Optical Bandwidth  $[Hz]$

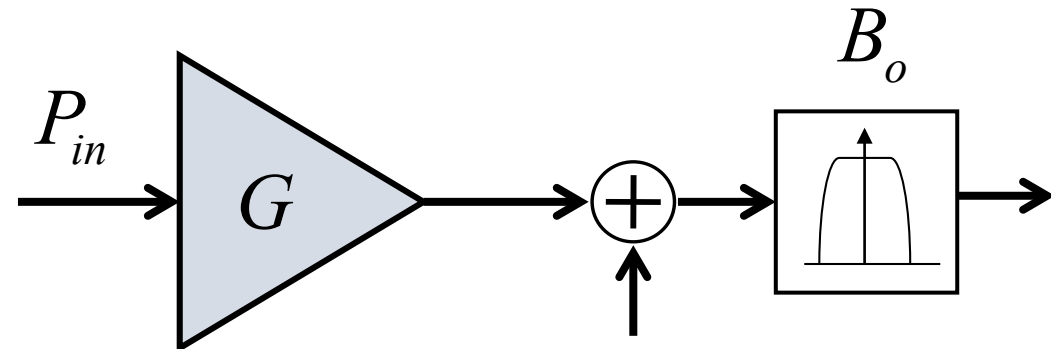
# Optical Signal to Noise Ratio (OSNR)

$$OSNR_{B_{ref}} = \frac{P_s}{S_{ase}(f_c) B_{ref}} \rightarrow P_{ase} = S_{ase}(f_c) B_o = \frac{P_s}{OSNR_{B_{ref}}} \frac{B_o}{B_{ref}}$$



$B_{ref}$ : Reference Bandwidth [Hz]  
 0.1 nm (12.5 GHz)  
 (Typical Value)

# ASE in Erbium-Doped Fiber Amps. (EDFA)



$$P_{out} = G(f_c) P_{in} \quad (\text{Signal})$$

$$P_{ase} \approx S_{ase}(f_c) B_o \quad (\text{Noise})$$

$$S_{ase} = hf (G(f) - 1) 2n_{sp}$$

$h$ : Planck's Constant [ $J \cdot s$ ]

$G$ : Amplifier's Gain

$n_{sp} \geq 1$ : Spontaneous Em. Factor

$QL$ : Quantum Limit

$$OSNR_{out} = \frac{\overbrace{GP_{in}}^{P_{out}}}{\underbrace{hf_c (G-1) 2n_{sp} B_{ref}}_{P_{ase}}} \approx \frac{P_{in}}{hf_c 2n_{sp} B_{ref}}$$

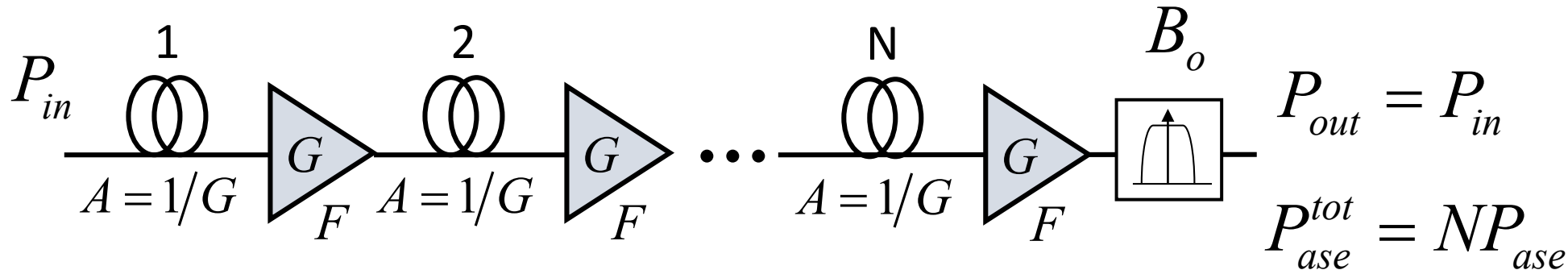
## Noise Figure

$$F \equiv \frac{OSNR_{in}}{OSNR_{out}} \approx \frac{\overbrace{P_{in} / hf_c \cdot B_{ref}}^{OSNR_{QL}}}{P_{in} / hf_c 2n_{sp} B_{ref}} = \underbrace{2n_{sp}}_{\geq 2}$$

## Noise Figure Measurement

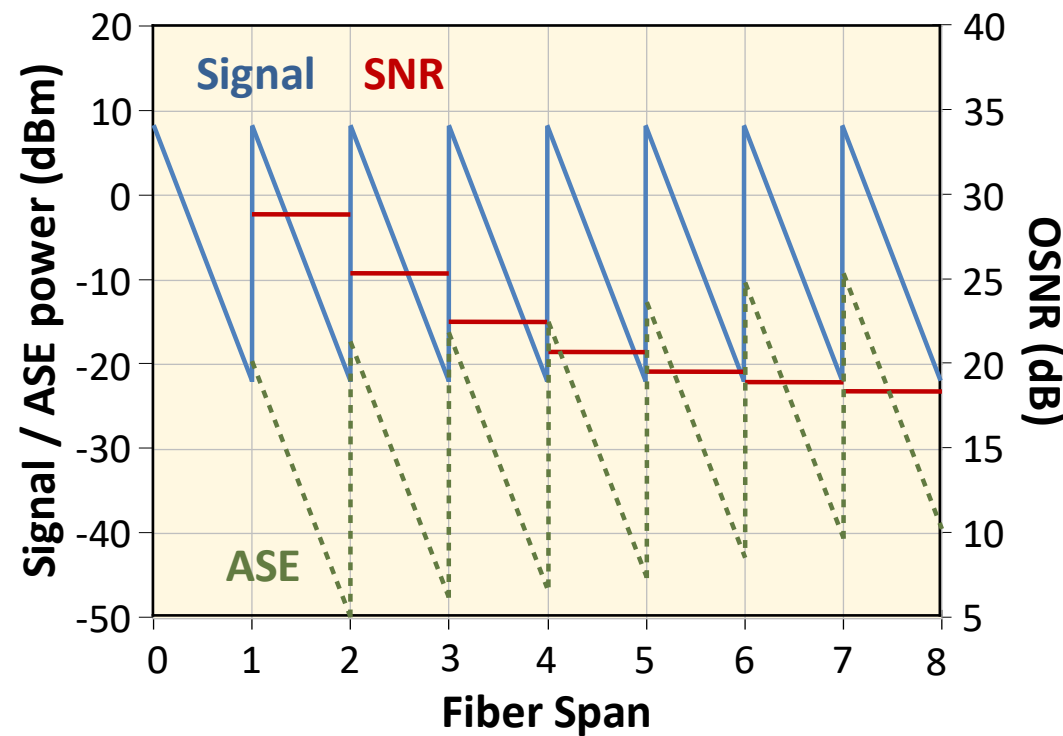
$$F_{dB} = P_{in}^{dBm} - OSNR_{out}^{dB} - \underbrace{10 \log_{10}(hf_c B_{ref})}_{\text{Quantum Noise } (-58dBm @ 0.1nm)}$$

# OSNR in an Link with EDFAs



$$OSNR_{link} = \frac{P_{in}}{NP_{ase}}$$

$$P_{ase} \approx hf_c (G - 1) F \cdot B_o$$





**BIT ERROR RATIO (BER) ESTIMATION**

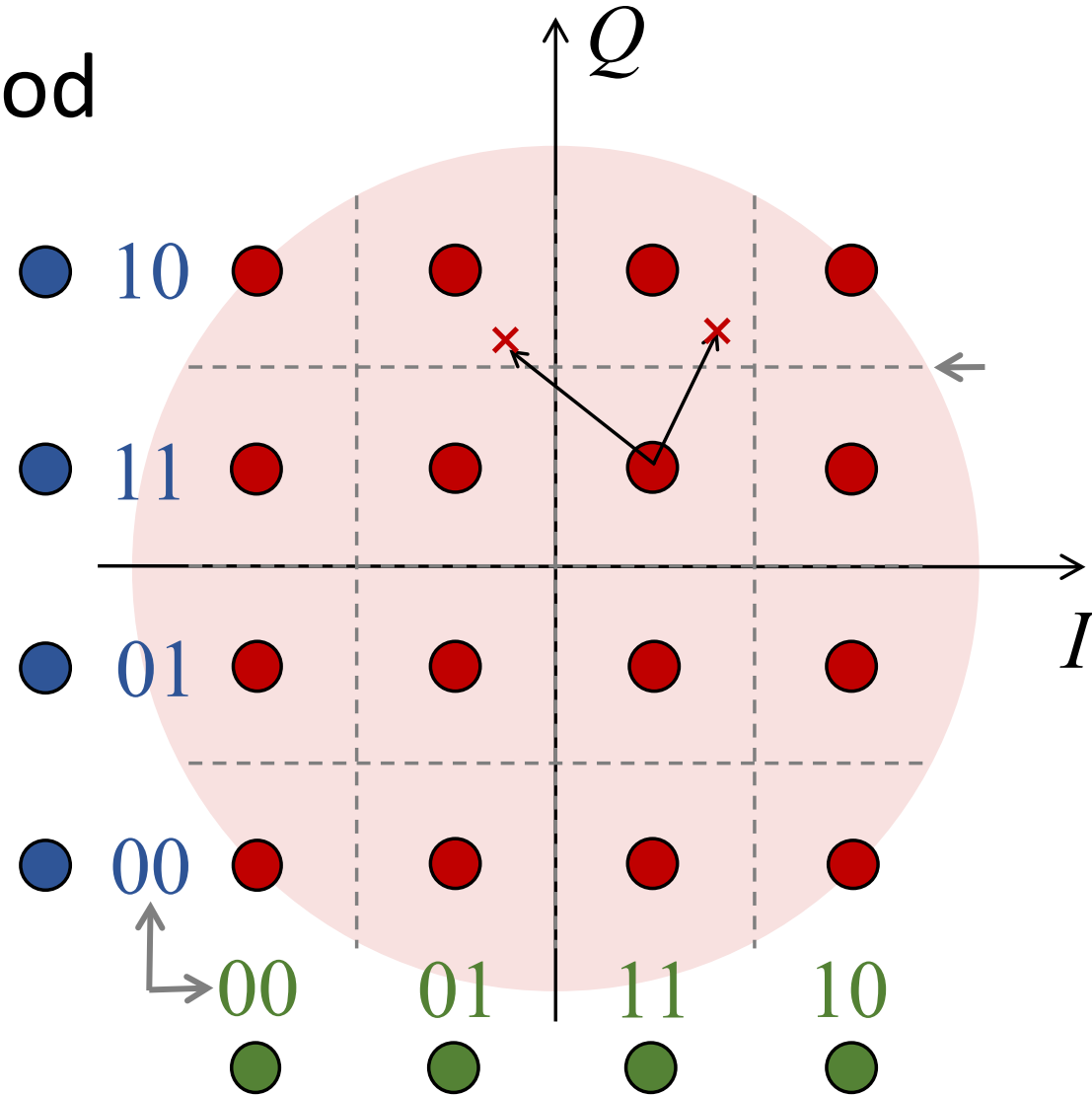
Monte Carlo Method (Error Count)

$$BER = \frac{\# Errors}{\# Bits}$$

$$BER_{ref} = 10^{-3} \rightarrow 10^{-9}$$

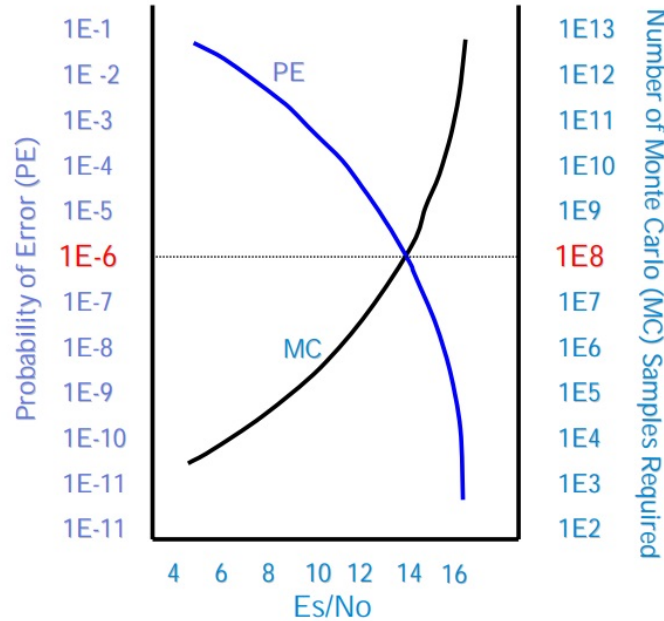
↑  
FEC codes

$$\underbrace{P(E)_{bit}}_{BER} \approx P(E)_{symbol}$$



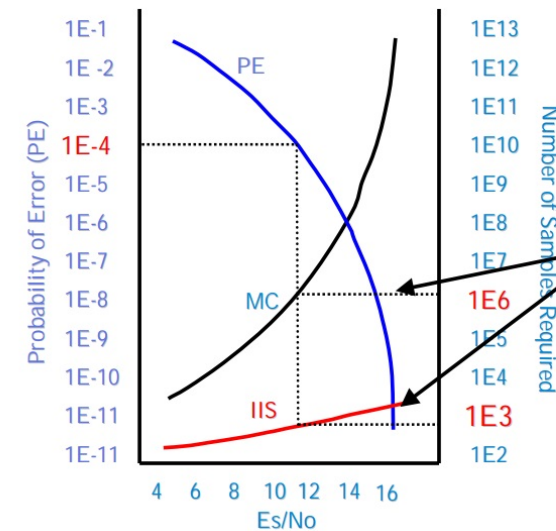
## Number of Required Samples

### Using Monte Carlo BER Estimation (99% Confidence)



### BER - Using Importance Sampling

How to save simulation time...



A PE of 1E-4 would require 1E6 samples for Monte Carlo vs. 1E3 samples for Importance Sampling

References

- Dingqiang Lu and Kung Yao, "Improved Importance Sampling Technique for Efficient Simulation of Digital Communication Systems", IEEE Journal on Selected Areas in Communications, Vol. 6, No. 1, pp. 67-75, January 1988
- Dingqiang Lu and Kung Yao, "Estimation Variance Bounds Of Importance Sampling Simulations in Digital Communication Systems", IEEE Transactions on Communications, Vol. 39, No. 10, October, 1991



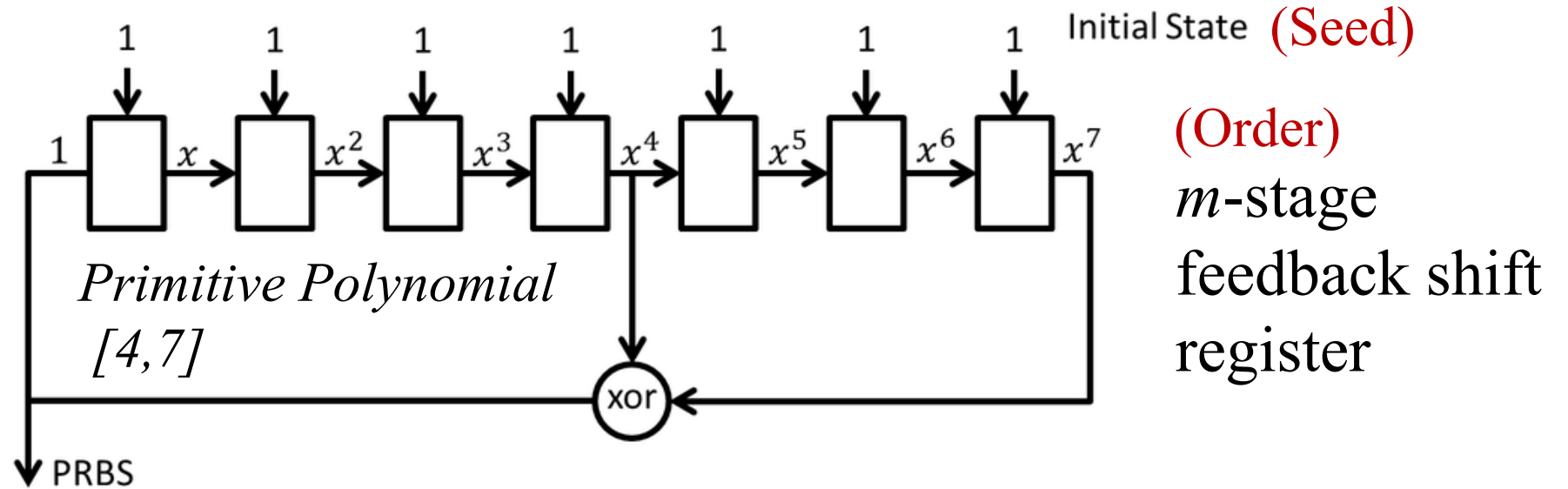
Agilent EEsos EDA

<http://literature.cdn.keysight.com/litweb/pdf/5989-9111EN.pdf>

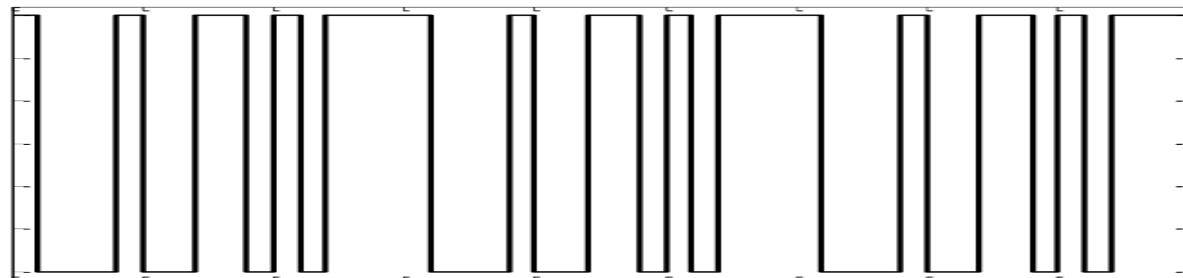
Presentation on Error Rate Simulations

**BIT ERROR RATIO (BER) ESTIMATION**

# Pseudo Random Bit Sequences (PRBS)



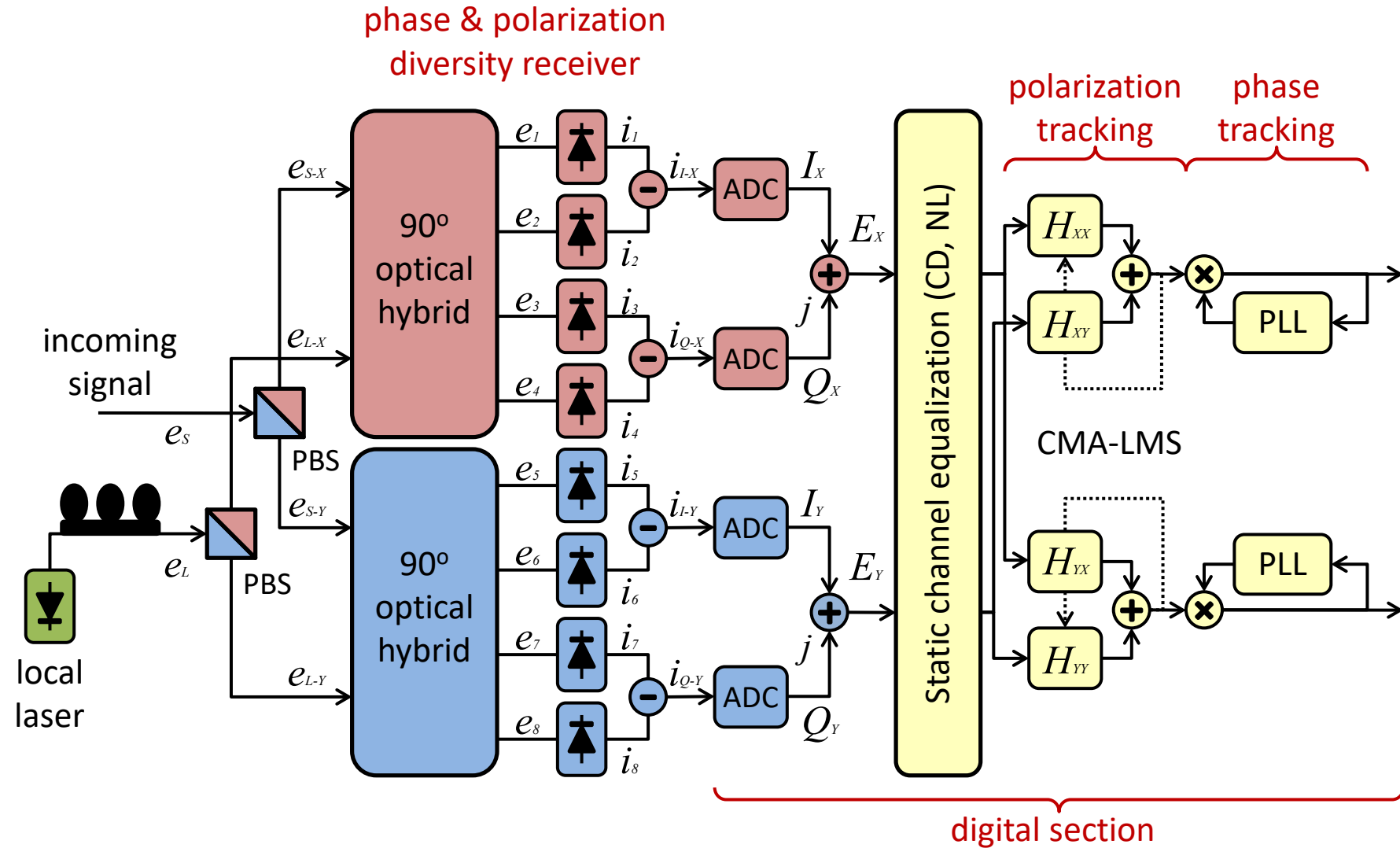
15 bits  $m = 4$



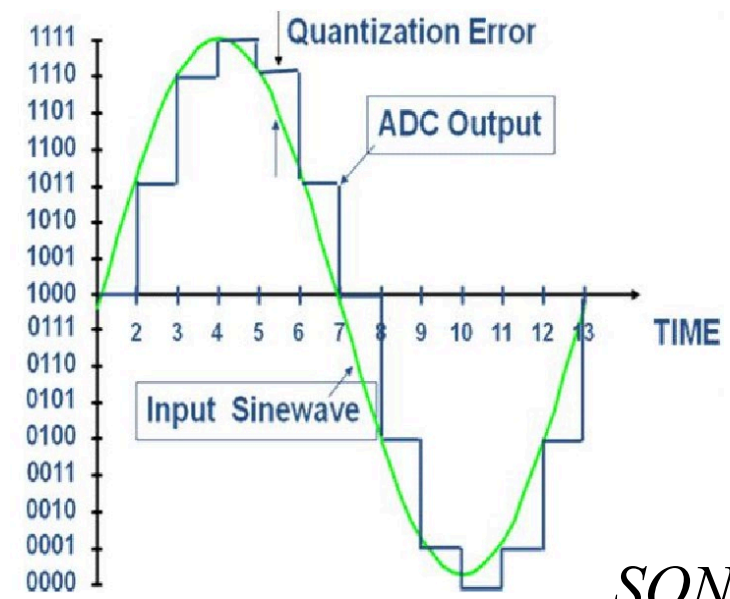
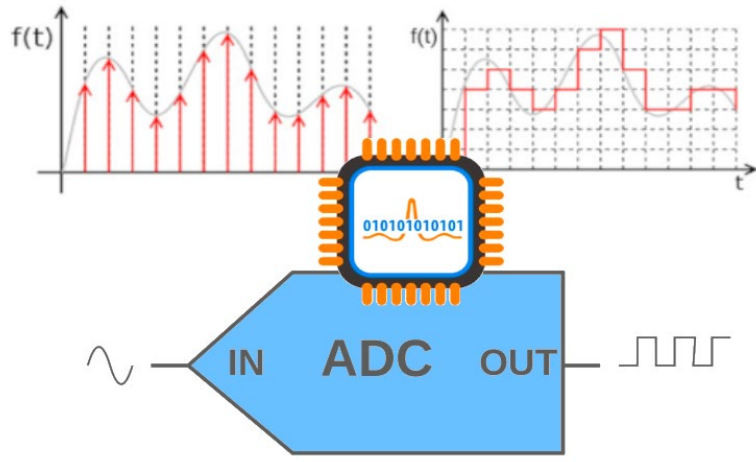
A maximal-length sequence generated will contain all but one  $m$ -bit combinations within a length of  $2^m - 1$  bits.

Michel C. Jeruchim et. al., "Simulation of Communication Systems", Kluwer Academic, Second Edition, 2002.

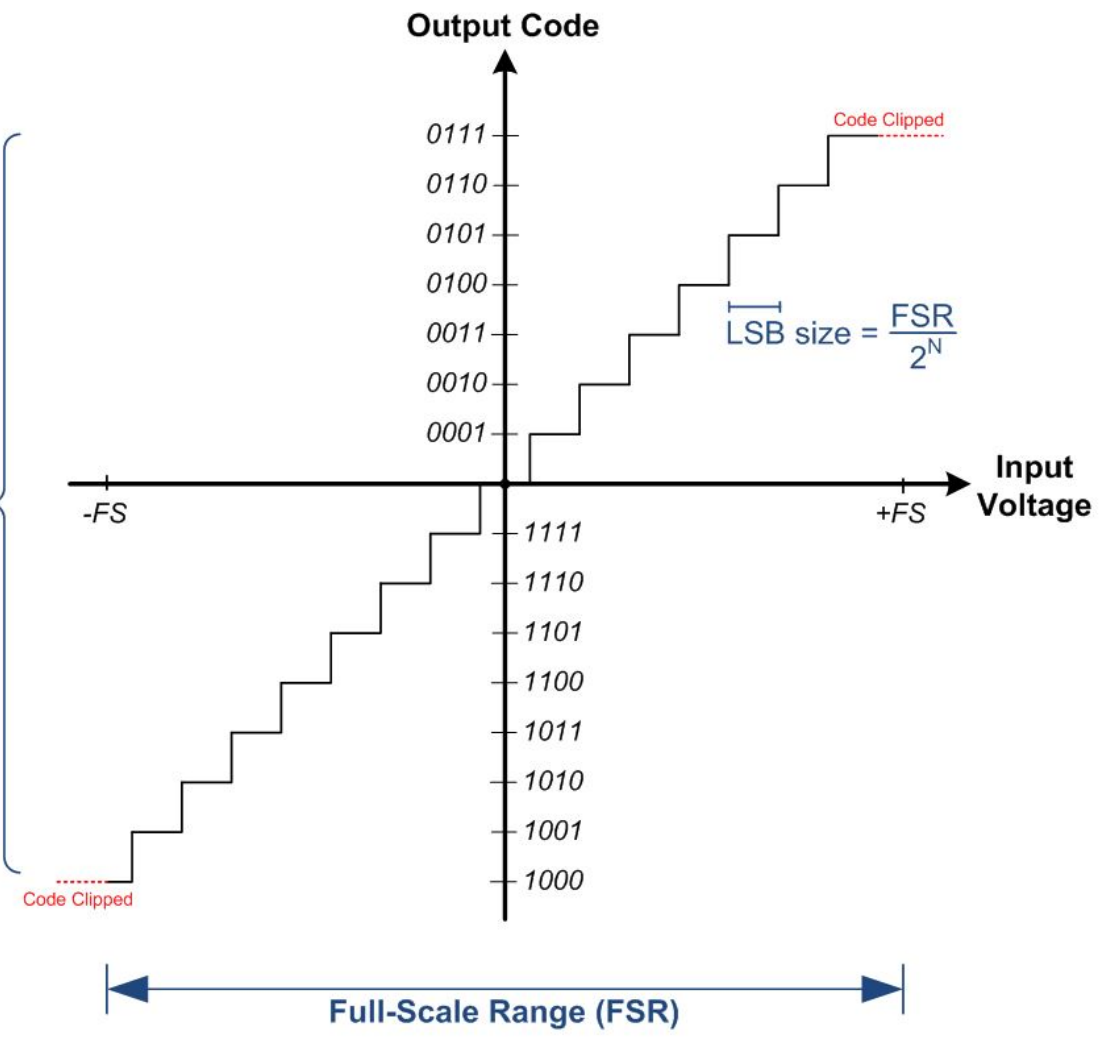
# DIGITAL COHERENT RECEIVER



# Analog to Digital Conversion



$2^N$  Codes



$$SQNR = 20 \log(2^N) \rightarrow SQNR_{dB} \approx 6N$$

# Chromatic Dispersion Compensation

Linear Regime  $\frac{\partial A}{\partial z} = j \frac{\beta_2}{2} \frac{\partial^2 A}{\partial t^2} \xrightarrow{FT} \frac{\partial \mathcal{A}}{\partial z} = -j\omega^2 \frac{\beta_2}{2} \mathcal{A}$

$$\mathcal{A}(z, \omega) = \underbrace{\mathcal{A}(0, \omega)}_{H(z, \omega)} e^{-j\omega^2 \frac{\beta_2}{2} z} \longrightarrow h(z, t) = \frac{1}{(j2\pi\beta_2 z)^{1/2}} e^{j\frac{t^2}{2\beta_2 z}}$$

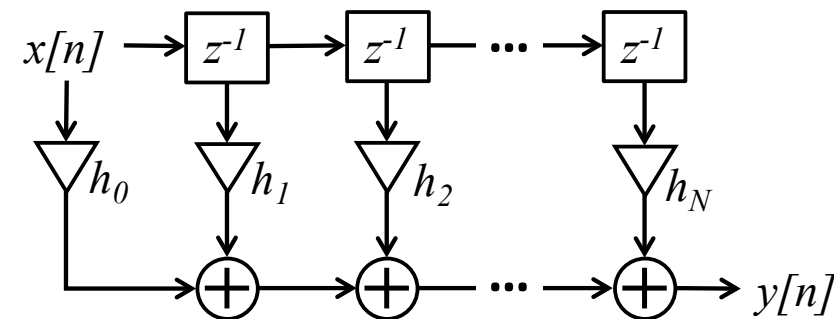
A signal sampled every  $T_{ADC}$  seconds can be recovered by applying a finite impulse response (FIR) filter to the signal with tap weights:

$$h[n] = \frac{1}{\sqrt{\rho}} e^{j\frac{\pi}{\rho} \left[ n - \frac{N-1}{2} \right]^2}$$

$$n \in [0, N-1]$$

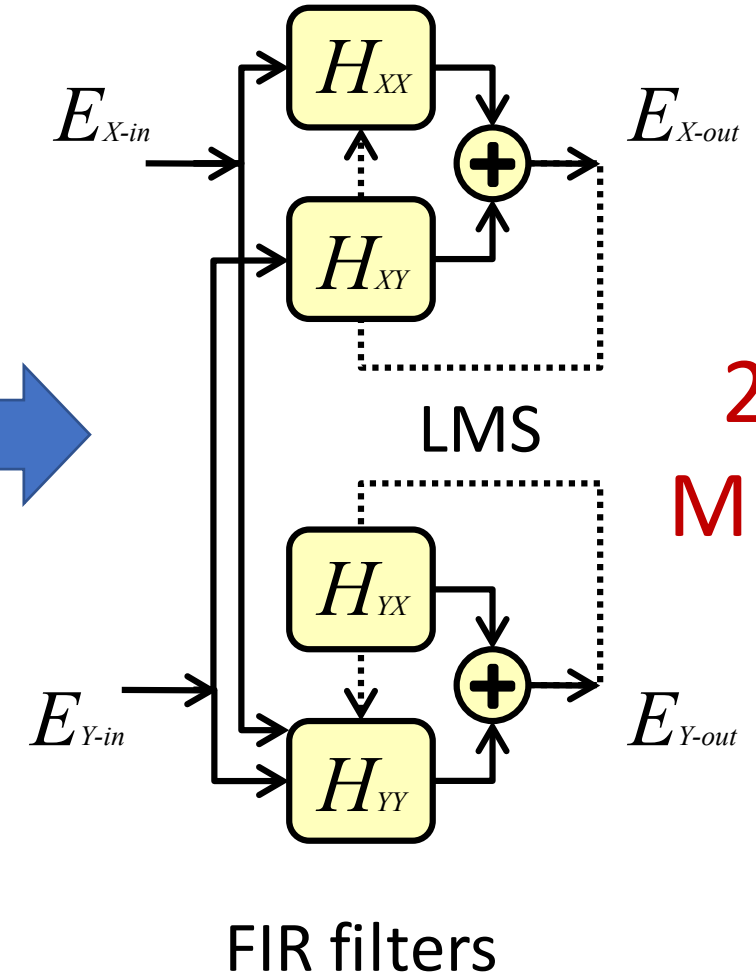
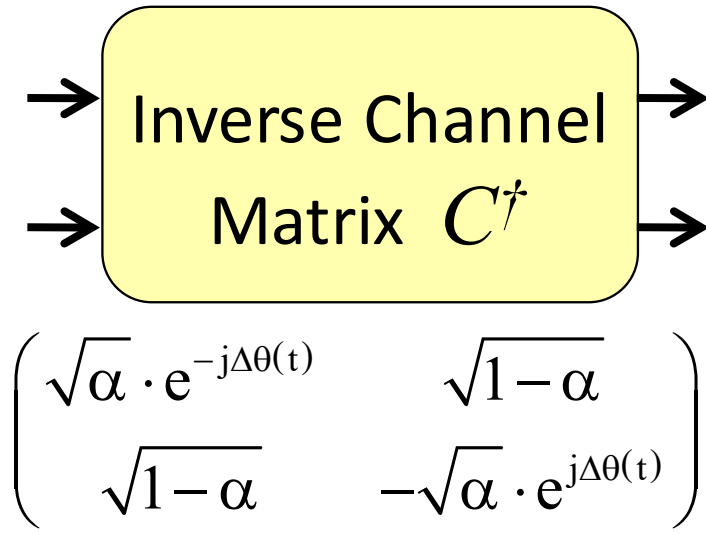
$$N = \lfloor |\rho| \rfloor$$

$$\rho = 2\pi\beta_2 \frac{L}{T_{ADC}^2}$$



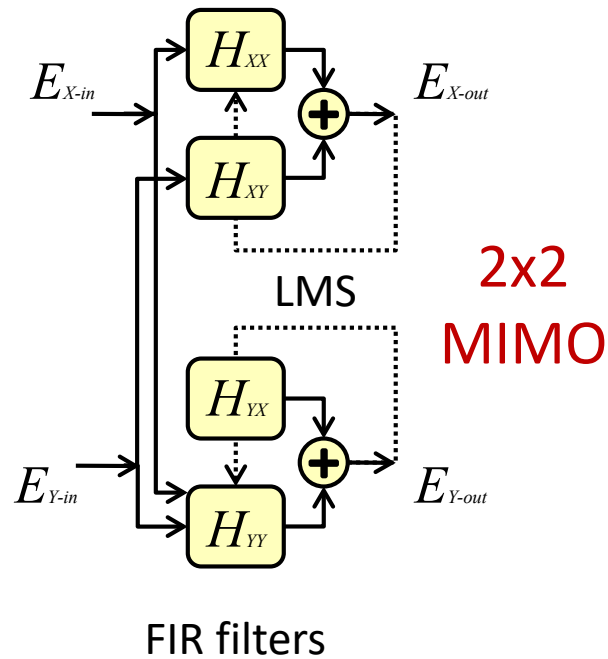
N: Number of Taps

# Polarization Tracking & PMD Compensation





# Least Mean Square (LMS) Algorithm

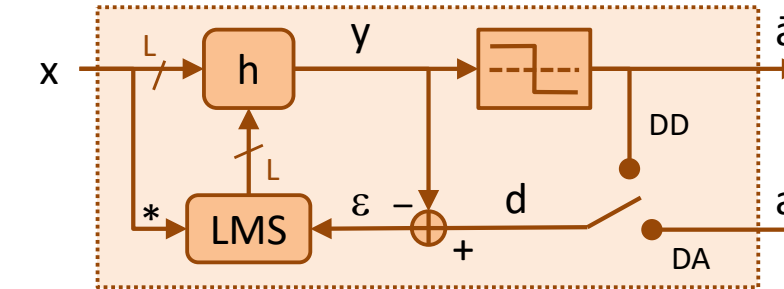


error function

$$\epsilon_X = d - A_{X-out}$$

$$\epsilon_Y = d - A_{Y-out}$$

d: decided data



filter updating mechanism

$$h_{XX}^{k+1} \mapsto h_{XX}^k + \mu \cdot \epsilon_X^k \cdot A_{X-in}^{*k}$$

$$h_{XY}^{k+1} \mapsto h_{XY}^k + \mu \cdot \epsilon_X^k \cdot A_{Y-in}^{*k}$$

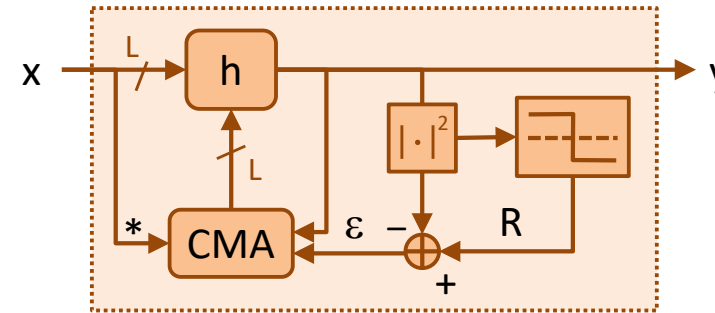
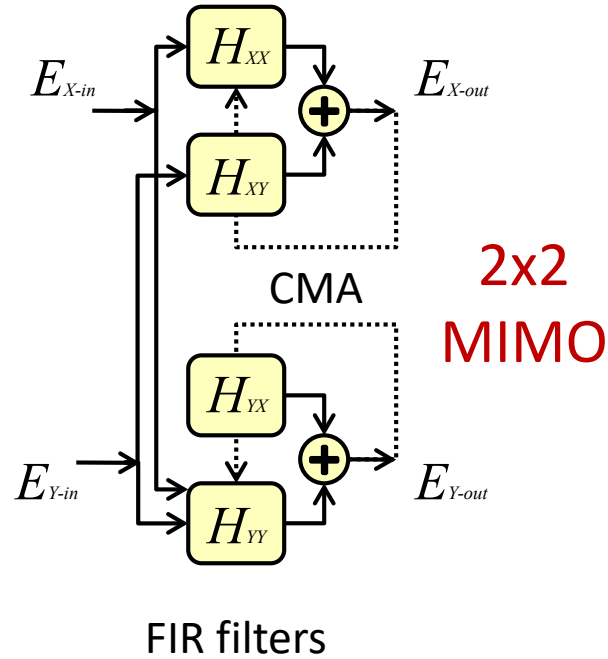
$$h_{YX}^{k+1} \mapsto h_{YX}^k + \mu \cdot \epsilon_Y^k \cdot A_{X-in}^{*k}$$

$$h_{YY}^{k+1} \mapsto h_{YY}^k + \mu \cdot \epsilon_Y^k \cdot A_{Y-in}^{*k}$$

$\mu$ : Convergence parameter ( $\sim 10^{-4}$ )

- Fast Convergence
- Training sequence required
- High computational cost
- High sensitivity to frequency and phase misalignment (within PLL)

# Constant Modulus Algorithm (CMA)



- Blind filter adaptation (no training sequence)
- Robust adaptive algorithm
- Independent of carrier frequency and phase (before PLL)
- Pre-convergence for QAM

## error function

$$\epsilon_X = (R^2 - |A_{X-out}|^2)$$

$$\epsilon_Y = (R^2 - |A_{Y-out}|^2)$$

R: circle radius

## filter updating mechanism

$$h_{XX}^{k+1} \mapsto h_{XX}^k + \mu \cdot \epsilon_X^k \cdot A_{X-in}^{*k} A_{X-out}^k$$

$$h_{XY}^{k+1} \mapsto h_{XY}^k + \mu \cdot \epsilon_X^k \cdot A_{Y-in}^{*k} A_{X-out}^k$$

$$h_{YX}^{k+1} \mapsto h_{YX}^k + \mu \cdot \epsilon_Y^k \cdot A_{X-in}^{*k} A_{Y-out}^k$$

$$h_{YY}^{k+1} \mapsto h_{YY}^k + \mu \cdot \epsilon_Y^k \cdot A_{Y-in}^{*k} A_{Y-out}^k$$

$\mu$ : Convergence parameter ( $\sim 10^{-2}$ )

# Constant Modulus Algorithm (CMA)

$$E_{\parallel} = \frac{R}{\sqrt{2}} |A_S^{\parallel}(t)| |A_L| \left( \frac{\sqrt{\alpha} \cdot e^{j\Delta\theta(t)}}{\sqrt{1-\alpha}} \right) e^{j\left(\omega_{FI}t + \overbrace{\theta_S^{\parallel}(t) - \theta_L(t)}^{\theta_{\parallel}(t)}\right)}$$

$$E_{\perp} = \frac{R}{\sqrt{2}} |A_S^{\perp}(t)| |A_L| \left( \frac{\sqrt{1-\alpha}}{-\sqrt{\alpha} \cdot e^{-j\Delta\theta(t)}} \right) e^{j\left(\omega_{FI}t + \overbrace{\theta_S^{\perp}(t) - \theta_L(t)}^{\theta_{\perp}(t)}\right)}$$

$$\begin{aligned} |E_X^{\parallel} + E_X^{\perp}|^2 &= \left| R\sqrt{\frac{\alpha}{2}} |A_S^{\parallel}(t)| |A_L| e^{j\left(\omega_{FI}t + \overbrace{\theta_S^{\parallel}(t) - \theta_L(t) + \Delta\theta(t)}^{\theta_{\parallel}(t)}\right)} + R\sqrt{\frac{1-\alpha}{2}} |A_S^{\perp}(t)| |A_L| e^{j\left(\omega_{FI}t + \overbrace{\theta_S^{\perp}(t) - \theta_L(t)}^{\theta_{\perp}(t)}\right)} \right|^2 = \\ &= \frac{R^2}{2} \left( \alpha |A_S^{\parallel}(t)|^2 + (1-\alpha) |A_S^{\perp}(t)|^2 \right) |A_L|^2 + R^2 \sqrt{\alpha} \sqrt{1-\alpha} |A_S^{\parallel}(t)| |A_S^{\perp}(t)| |A_L|^2 \cos(\theta_S^{\parallel}(t) - \theta_S^{\perp}(t) + \Delta\theta(t)) \end{aligned}$$

$$\begin{aligned} |E_Y^{\parallel} + E_Y^{\perp}|^2 &= \left| R\sqrt{\frac{1-\alpha}{2}} |A_S^{\parallel}(t)| |A_L| e^{j\left(\omega_{FI}t + \overbrace{\theta_S^{\parallel}(t) - \theta_L(t)}^{\theta_{\parallel}(t)}\right)} - R\sqrt{\frac{\alpha}{2}} |A_S^{\perp}(t)| |A_L| e^{j\left(\omega_{FI}t + \overbrace{\theta_S^{\perp}(t) - \theta_L(t) - \Delta\theta(t)}^{\theta_{\perp}(t)}\right)} \right|^2 = \\ &= \frac{R^2}{2} \left( (1-\alpha) |A_S^{\parallel}(t)|^2 + \alpha |A_S^{\perp}(t)|^2 \right) |A_L|^2 - R^2 \sqrt{\alpha} \sqrt{1-\alpha} |A_S^{\parallel}(t)| |A_S^{\perp}(t)| |A_L|^2 \cos(\theta_S^{\parallel}(t) - \theta_S^{\perp}(t) + \Delta\theta(t)) \end{aligned}$$

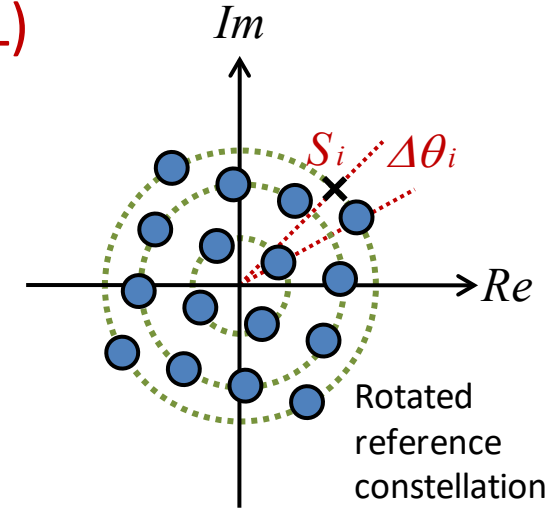
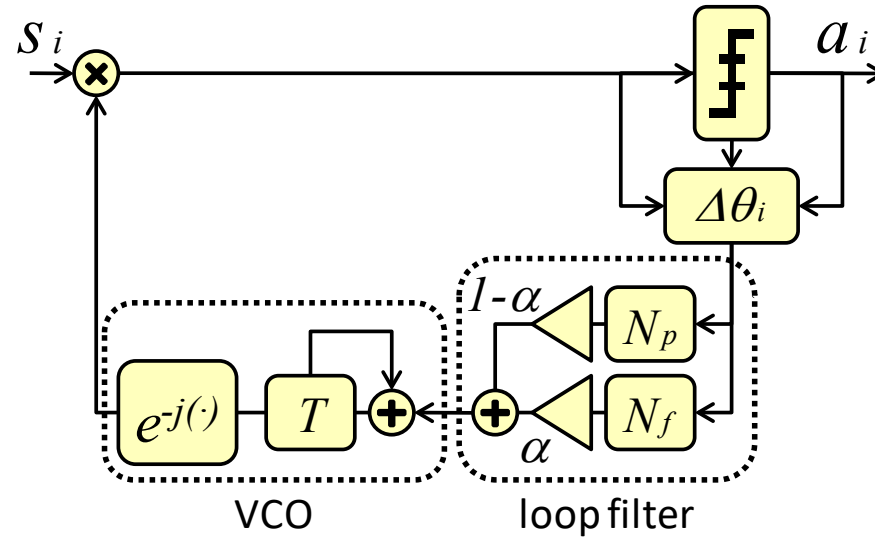
$$|A_S^{\parallel}(t)|^2 = |A_S^{\perp}(t)|^2 \quad \leftarrow \text{QPSK}$$

$$|E_X^{\parallel} + E_X^{\perp}|^2 = \frac{R^2}{2} |A_S(t)|^2 |A_L|^2 + R^2 \sqrt{\alpha} \sqrt{1-\alpha} |A_S(t)|^2 |A_L|^2 \cos(\theta_S^{\parallel}(t) - \theta_S^{\perp}(t) + \Delta\theta(t))$$

$$|E_Y^{\parallel} + E_Y^{\perp}|^2 = \frac{R^2}{2} |A_S(t)|^2 |A_L|^2 - R^2 \sqrt{\alpha} \sqrt{1-\alpha} |A_S(t)|^2 |A_L|^2 \cos(\theta_S^{\parallel}(t) - \theta_S^{\perp}(t) + \Delta\theta(t))$$

# Frequency & Phase Estimation

## Decision-Directed Phase-Locked Loop (DD-PLL)



When LMS is used, PLL is placed inside the LMS loop increasing the system's instability. CMA avoids this situation.

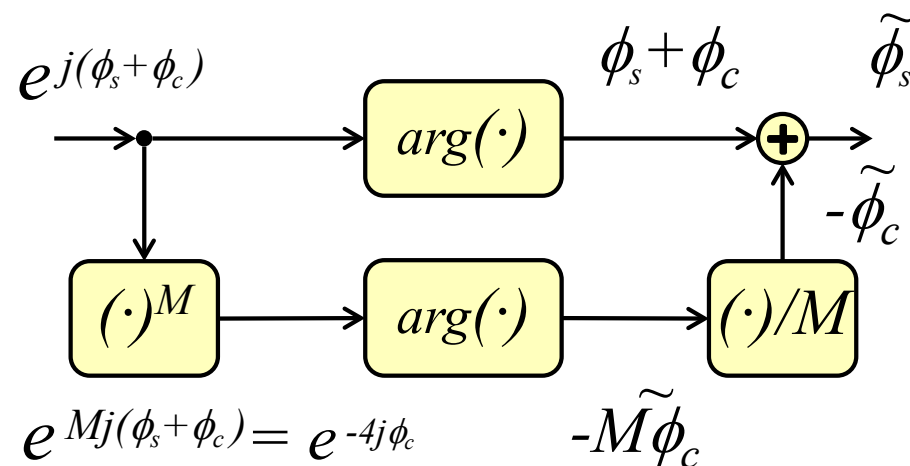
$\alpha \sim 0.95$   
 $N_f \sim 1000$  samples  
 $N_p \sim 10$  samples

Reference constellation steps ( $\pi/20$ ) following the smallest mean-square error (Maximum-Likelihood)

- Constellation spinning at IF
- Constellation wiggling due to phase noise
- Constellation rotated due to phase uncertainty
- QAM grid decision
- Gray code
- Short training sequence

# Viterbi & Viterbi Algorithm

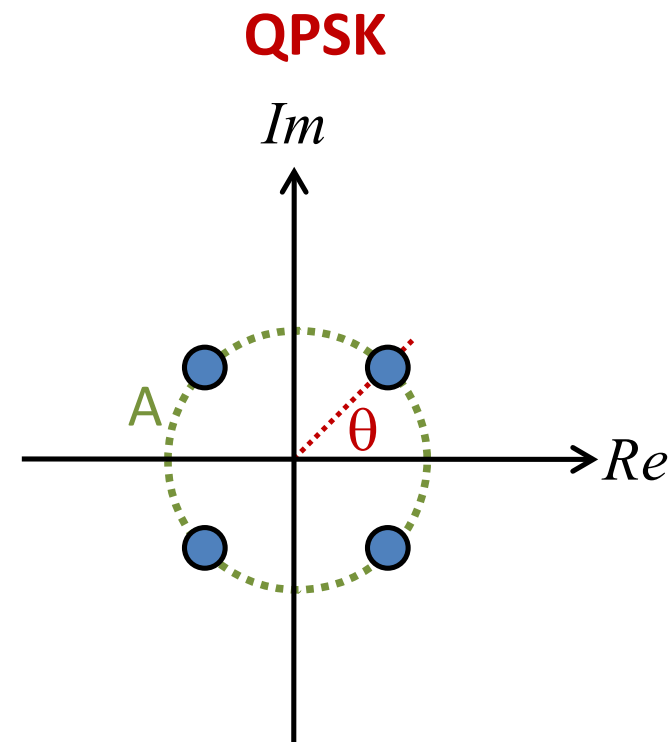
## Viterbi & Viterbi Algorithm



PSK

$$\phi_s = \frac{\pi}{M} (2m + 1) \quad , m = \{0..M\}$$

- Very simple implementation
- Works only for QPSK
- Multi-modulus variations for QAM



$$\phi_s = \left\{ \frac{\pi}{4}, \frac{3\pi}{4}, \frac{5\pi}{4}, \frac{7\pi}{4} \right\}$$