

Active Content Fingerprinting: a marriage of passive content fingerprinting and digital watermarking

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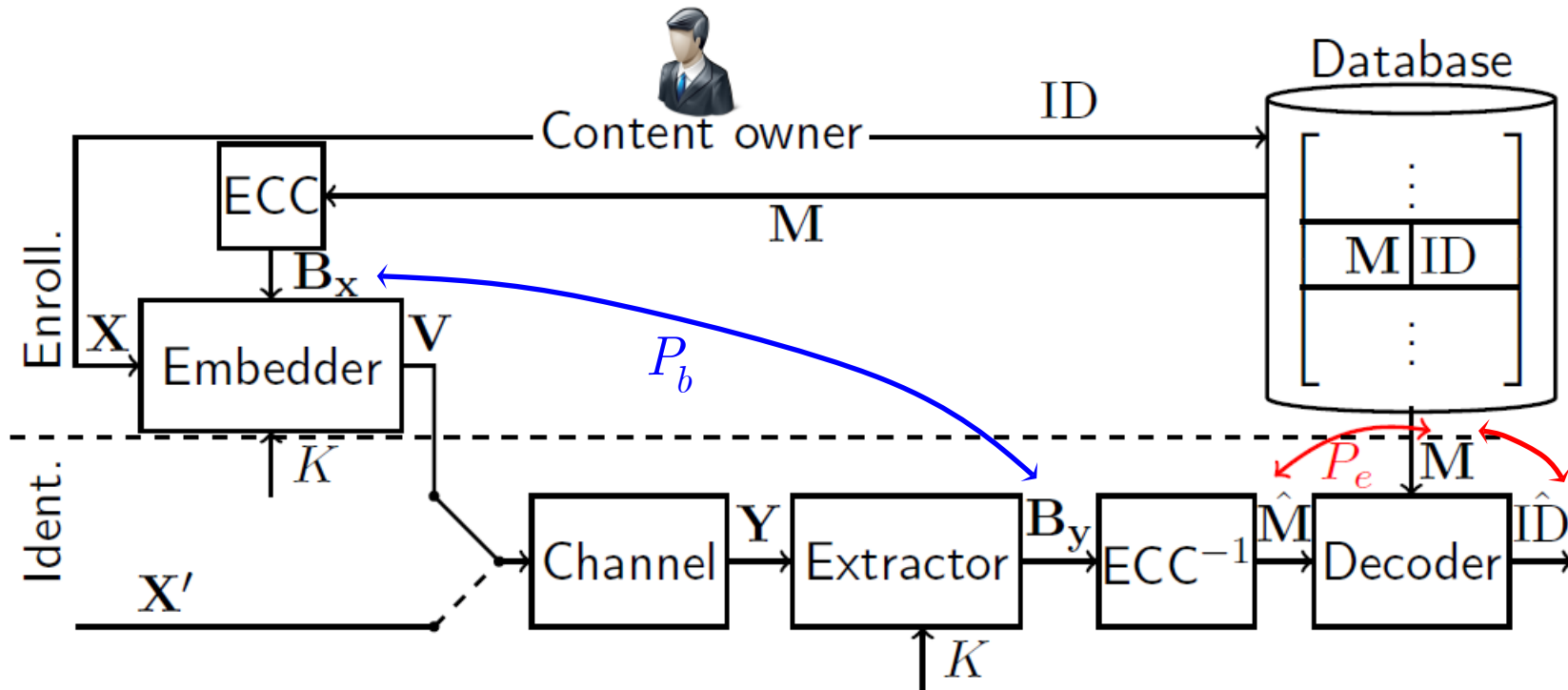
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Outline

- 1 Introduction: Digital watermarking and passive content fingerprinting
- 2 Active content fingerprinting
- 3 Simulation results
- 4 Conclusions

Traditional approaches: Digital watermarking

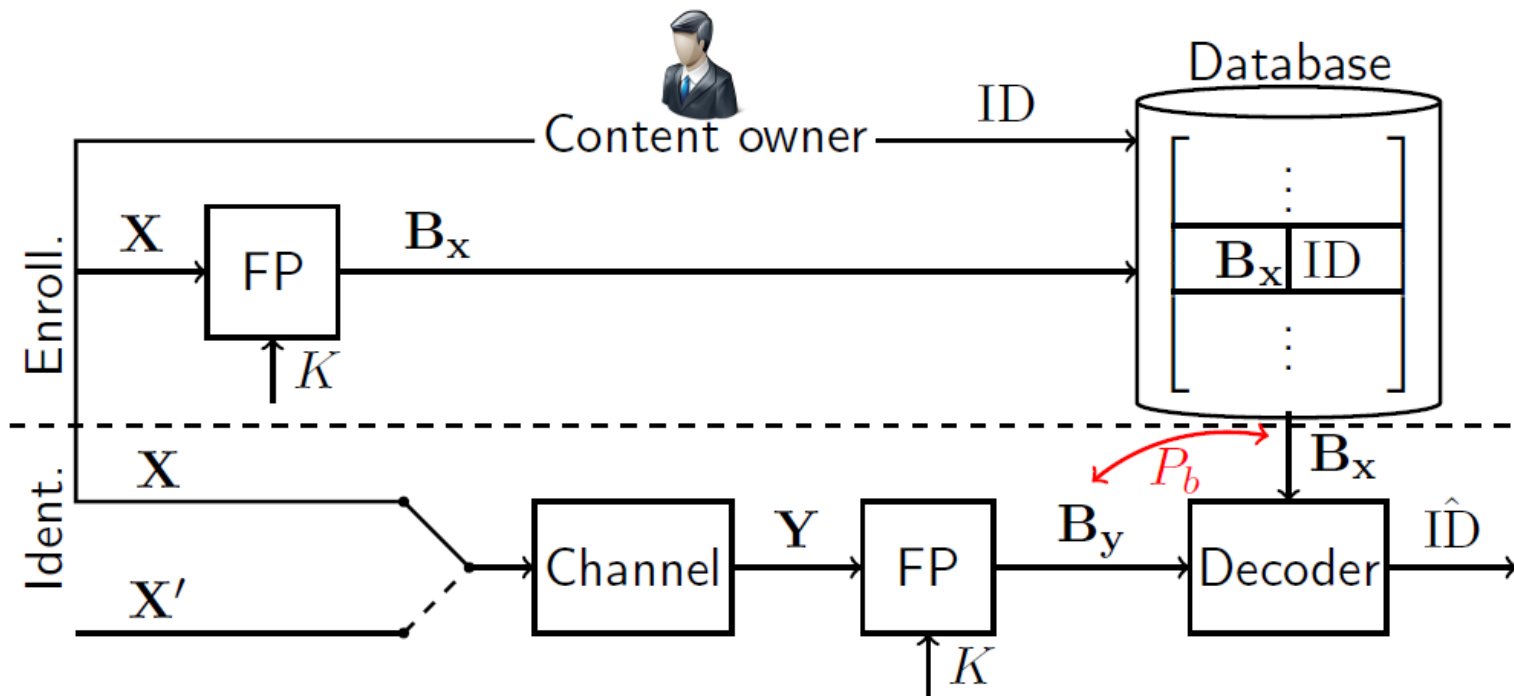
Digital Watermarking (DWM):



- embed host-independent message: $ID \rightarrow M$
- host interference
- distortions due to watermark embedding and host interference cancellation
- structured codes ($P_e \approx 0$) \Rightarrow low complexity BUT security concerns

Traditional approaches: Passive content fingerprinting

Passive Content Fingerprinting (pCFP):



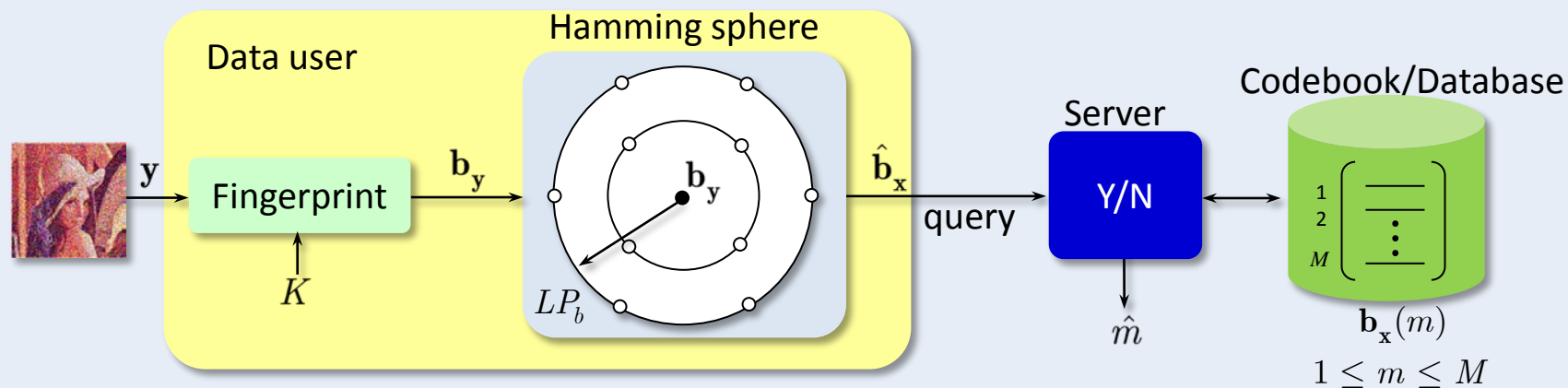
- no modifications
- random codes \Rightarrow exhaustive search \Rightarrow high complexity
- high probability of average bit error $P_b \Rightarrow$ low performance

Complexity of fingerprint based identification (Cont.)

Hamming sphere decoding

Observation: the most likely codewords $\mathbf{b}_x(\hat{m})$ are within Hamming sphere of radius γL around \mathbf{b}_y .

Identification = codeword verification [S. Voloshynovskiy et al, ITW2010, Dublin]

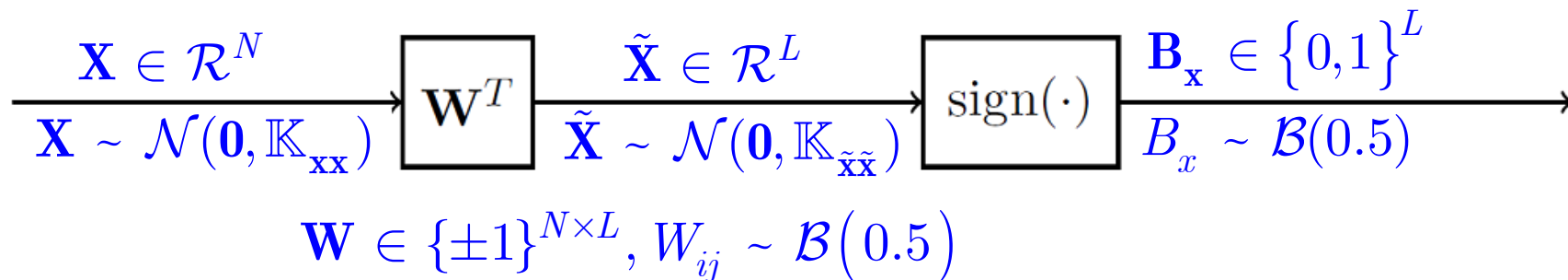


Remarks

- Complexity depends on “data quality”: $\mathcal{O}\left(L2^{LH_2(P_b)}\right)$
- In limit $P_b \rightarrow 0$, complexity $\Rightarrow \mathcal{O}(L)$ (noiseless hashing)

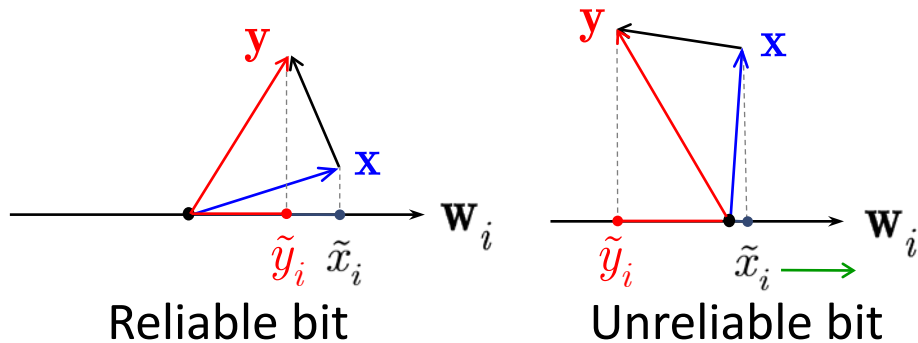
pCFP: factors determining performance

Fingerprinting (FP) block based on random projections:



Additive White Gaussian Noise (AWGN): $\mathbf{y} = \mathbf{x} + \mathbf{z}$

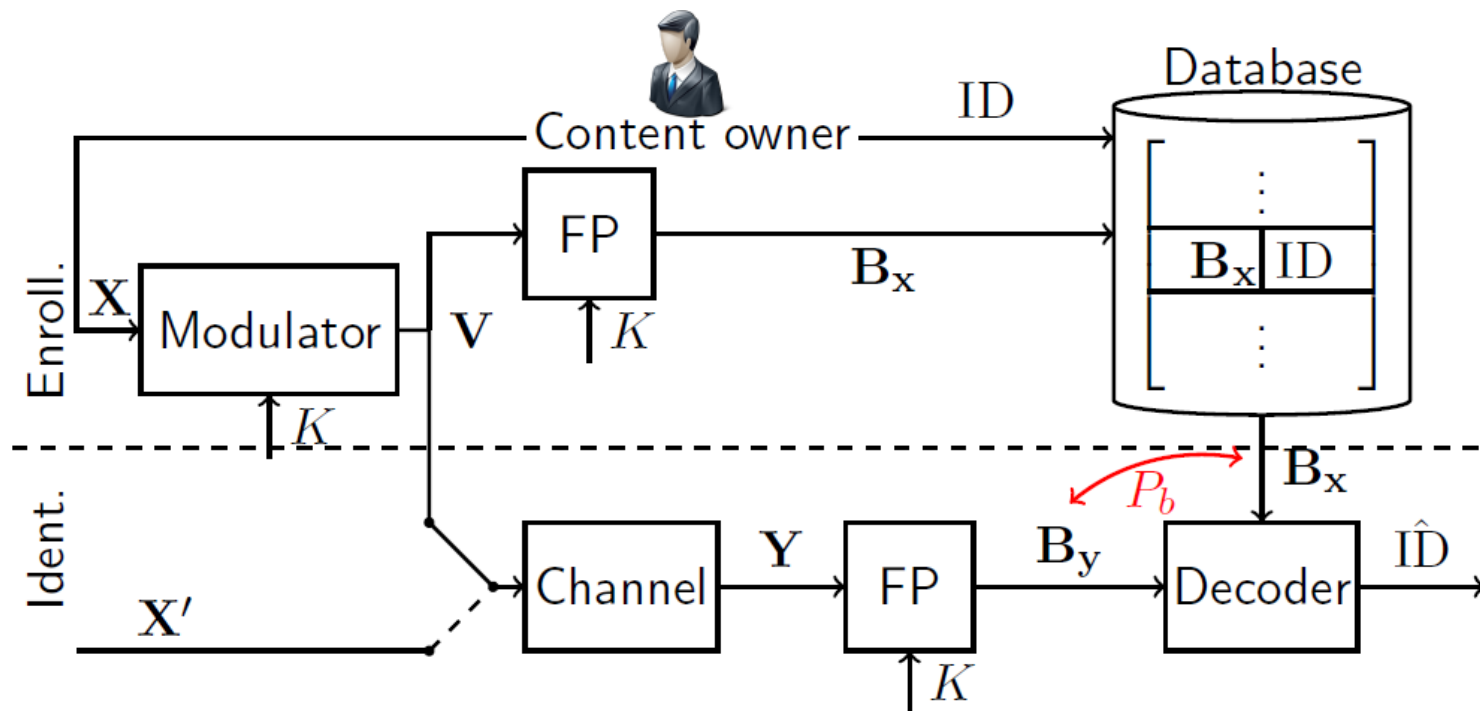
$$P_{b|\tilde{x}} = Q\left(\frac{|\tilde{x}|}{\sigma_Z}\right), P_{b-\text{pCFP}} = E[P_{b|\tilde{x}}] = \frac{1}{\pi} \arccos(\rho_{\tilde{X}\tilde{Y}})$$



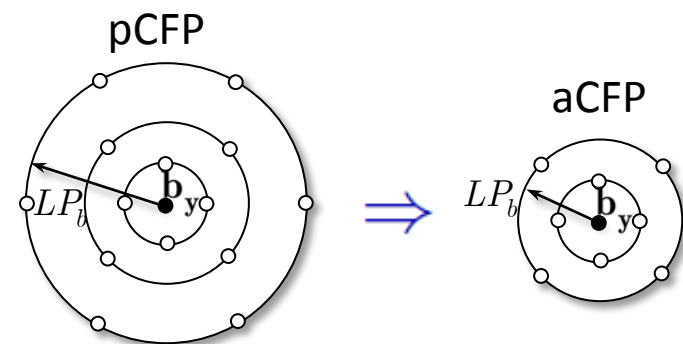
Can one increase $|\tilde{x}|$ by purpose?

Active Content Fingerprint (aCFP)

Proposed Approach:



- No message will be embedded (= pCFP)
- Content is modified
- Main goal: $P_b \rightarrow 0 \Rightarrow$ low search complexity



How it works?

modify images to make features (more) robust

- orthogonal expansion

$$\begin{cases} \tilde{x}_i &= \mathbf{w}_i^T \mathbf{x}, \quad 1 \leq i \leq N, \\ \mathbf{x} &= \sum_{i=1}^N \tilde{x}_i \mathbf{w}_i = \sum_{i \in \mathcal{K}} \tilde{x}_i \mathbf{w}_i + \sum_{i \notin \mathcal{K}} \tilde{x}_i \mathbf{w}_i \end{cases}$$

$$\mathcal{K} = \{i_1, i_2, \dots, i_L\}$$

- modified content

$$\mathbf{v} = \sum_{i \in \mathcal{K}} \varphi_i(\tilde{x}_i) \mathbf{w}_i + \sum_{i \notin \mathcal{K}} \tilde{x}_i \mathbf{w}_i,$$

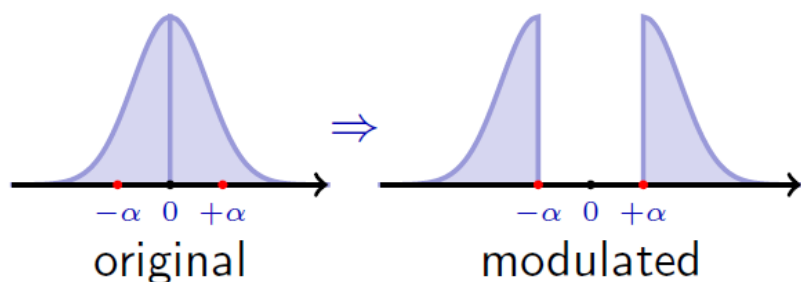
where $\varphi_i(\cdot), i \in \mathcal{K}$ is a modulation function, which can be the same for all i .

Additive aCFP (AddaCFP)

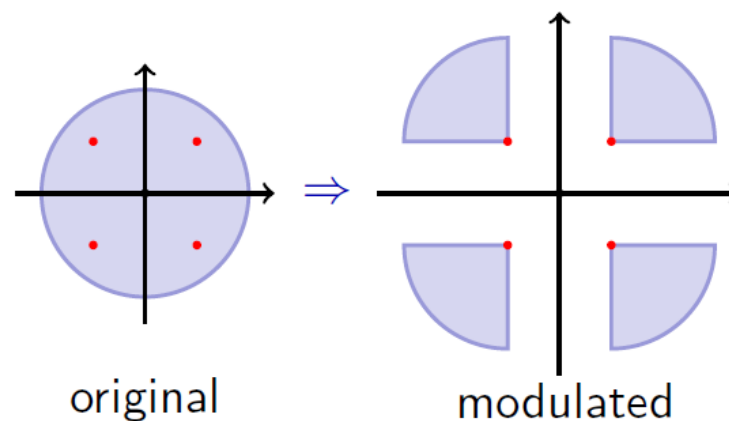
AddaCFP modulation function

$$\varphi_A(\tilde{x}_i) = \tilde{x}_i + \alpha \text{sign}(\tilde{x}_i), \forall i \in \mathcal{K}$$

One dimensional:



Two dimensional:



- Bit error probability (AWGN):

$$P_{b-\text{AddaCFP}} \leq \exp\left(-\frac{\alpha^2}{2\sigma_Z^2}\right) P_{b-\text{pCFP}}$$

With distortion

$$D_A = \frac{L}{N} \alpha^2, L = |\mathcal{K}|$$

Additive aCFP (AddaCFP): link to additive DWM

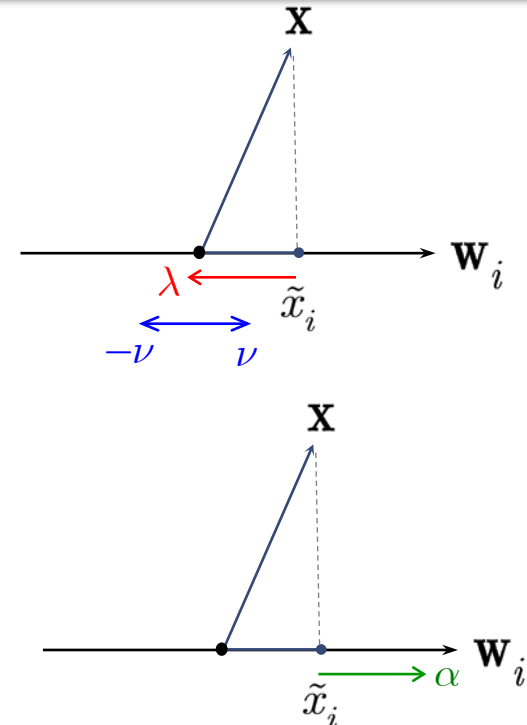
Link to improved spread spectrum (ISS) watermarking

$$\mathbf{v} = \mathbf{x} - \underbrace{\lambda \sum_{i \in \mathcal{K}} \tilde{x}_i \mathbf{w}_i}_{\text{Interference cancellation}} + \underbrace{\nu \sum_{i \in \mathcal{K}} (-1)^{m_i} \mathbf{w}_i}_{\text{DWM embedding}}, \forall i \in \mathcal{K}, m_i \in \{0, 1\}$$

- Two types of distortions due to:
 - host interference cancellation
 - WM embedding

$$\mathbf{v} = \mathbf{x} + \underbrace{\alpha \sum_{i \in \mathcal{K}} \text{sign}(\tilde{x}_i) \mathbf{w}_i}_{\text{AddaCFP content modulation}}$$

≡ Coefficient magnitude increase

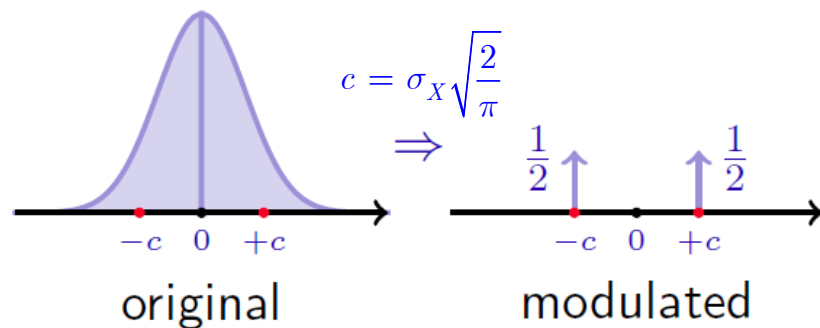


Quantization based aCFP (QbaCFP)

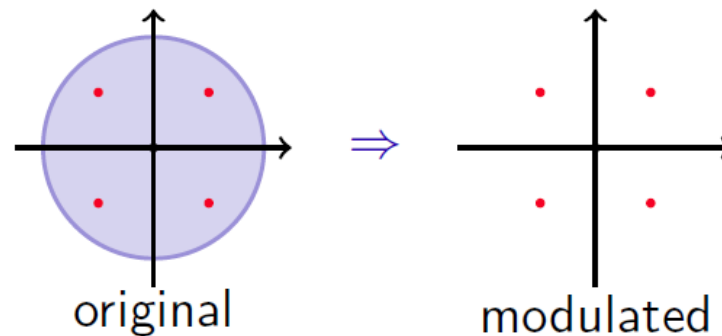
QbaCFP modulation function

$$\varphi_Q(\tilde{x}_i) = c \text{sign}(\tilde{x}_i), \forall i \in \mathcal{K}$$

One dimensional:



Two dimensional:



- Bit error probability (AWGN):

$$P_{b-QbaCFP} = Q\left(\frac{c}{\sigma_Z}\right) \stackrel{\min D_Q}{\Rightarrow} P_{b-QbaCFP} = Q\left(\frac{\sigma_X}{\sigma_Z} \sqrt{\frac{2}{\pi}}\right)$$

Quantization based aCFP (QbaCFP): link to QIM

Link to spread-transform dither modulation (ST-DM) watermarking

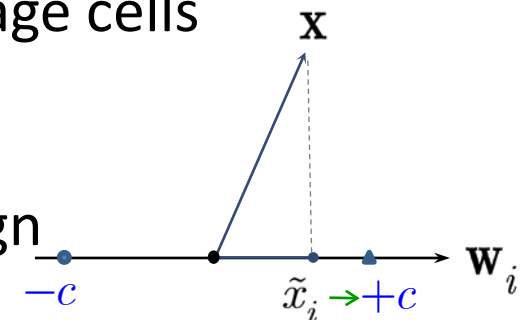
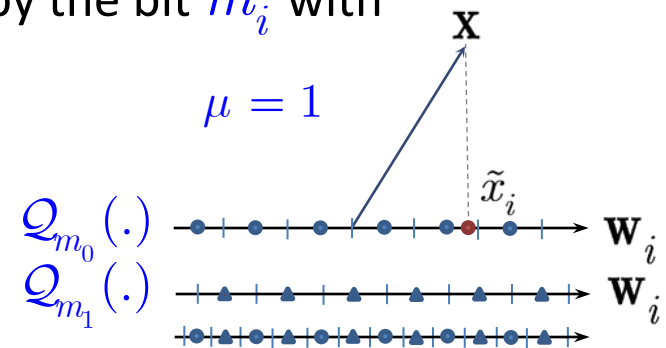
$$\mathbf{v} = \mathbf{x} + \mu \sum_{i \in \mathcal{K}} \left(\mathcal{Q}_{m_i}(\tilde{x}_i) - \tilde{x}_i \right) \mathbf{w}_i, \forall i \in \mathcal{K}, m_i \in \{0,1\}$$

where $\mathcal{Q}_{m_i}(\cdot)$ is a scalar quantizer, which is defined by the bit m_i with the centroids defined by

$$c_i = \Delta \mathcal{Z} + (-1)^{m_i} \frac{\Delta}{4}, \text{ for } m_i = 0,1$$

- The distortion is due to quantization
- The quantizers are message-dependent
- Errorless bit extraction, iff noise is within message cells

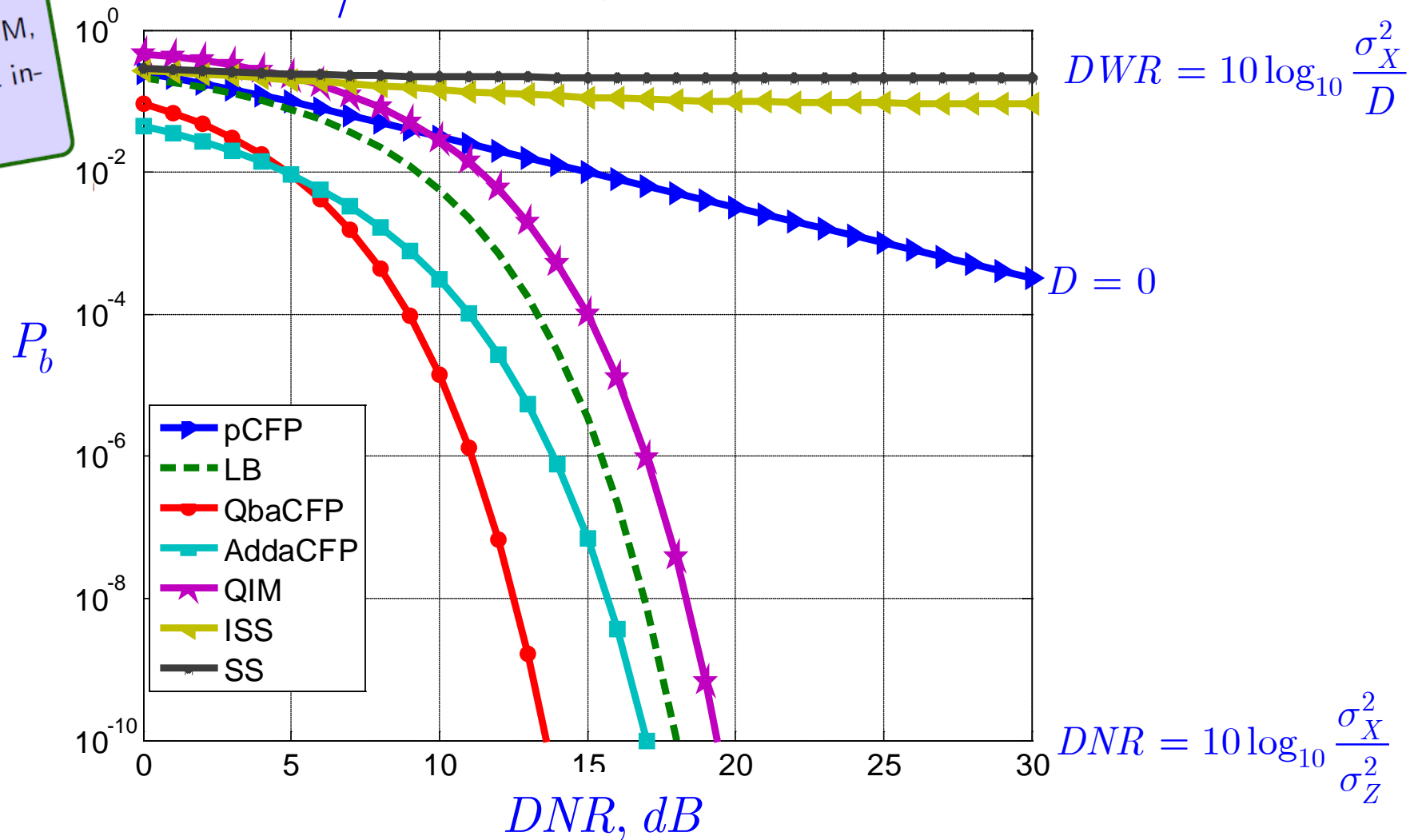
- The quantizer is message independent
- Errorless bit extraction, if noise does not flip sign
 \Rightarrow higher tolerance to noise distortions



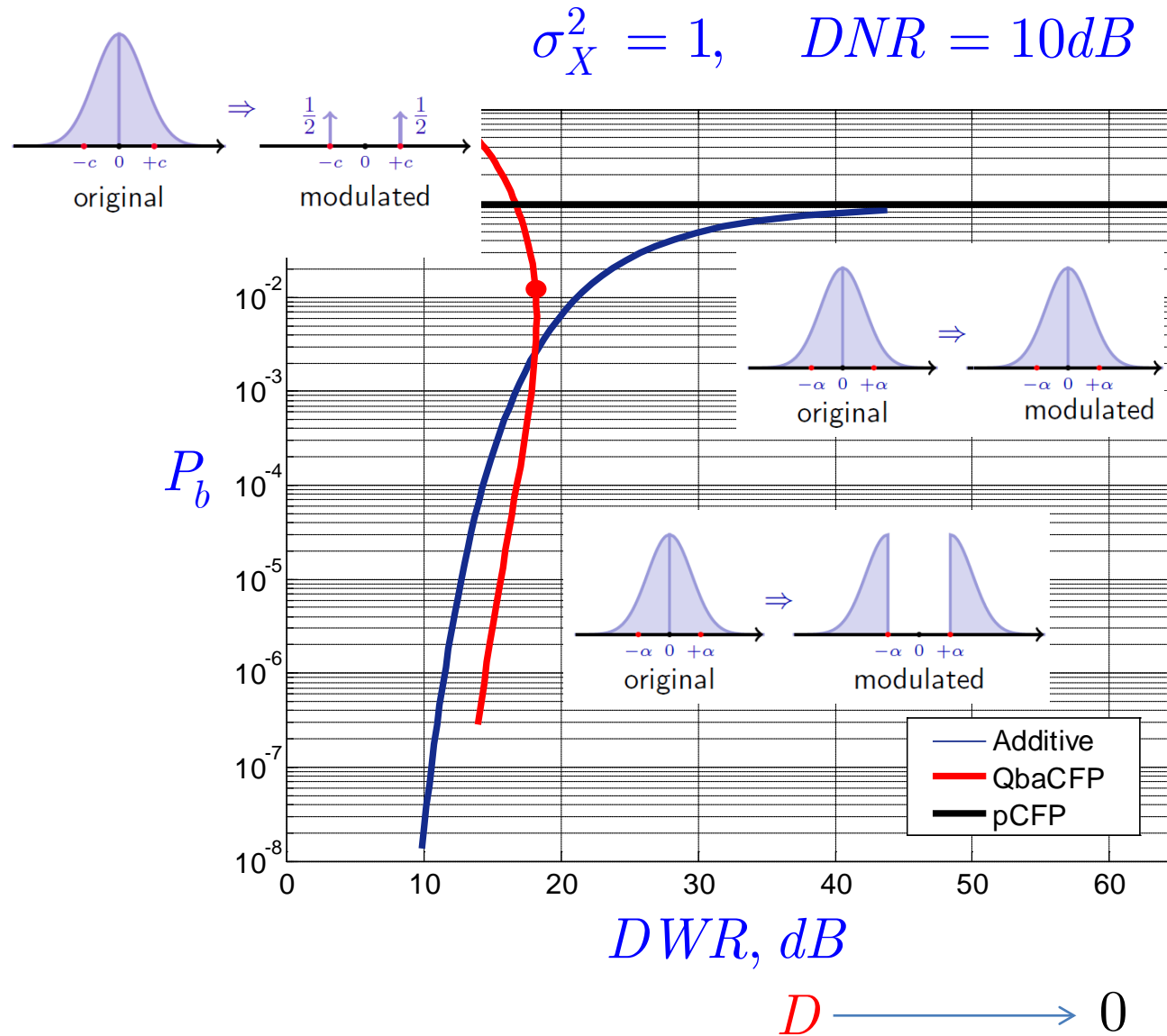
Analytical comparison

$$L/N = 0.016, DWR = 20dB$$

LB
Idealized DWM,
without host in-
terference.



Analytical comparison



Quantization based aCFP (QbaCFP)

Table: Comparison of BERs, $N = 768$, $L = 32$, $\alpha = 45$ and average PSNR=53dB (original to embedded image).

PSDR	Attack	Parameters	pCFP	AddaCFP
			P_{b-pCFP}	$P_{b-AddaCFP}$
AWGN		PSNR=5 dB	0.21	0.13
		PSNR = 10 dB	0.13	0.06
		PSNR = 15 dB	0.08	0.02
		PSNR = 20 dB	0.05	0.002
JPEG		QF = 1	0.11	0.04
		QF = 10	0.04	0.001
		QF = 25	0.01	0
Histeq			0.1	0.014

Feature vectors extracted from Uncompressed Colour Image Database (UCID), which consists of 1338 images of size 384 by 512.

Conclusions

- aCFP: combination of pCFP and DWM
- Improved performance in terms of bit error probability
- Potential huge gain for lower complexity based on BDD (“crypto hashing”)

Future extensions

- Optimal modulation to trade-off distortion-performance
- New insights for security and privacy
- Multidimensional and lattice extensions
- Application to physical object security
SNF project: Nano- and micro-structure identification based on controllable randomness