

Basics of Watermarking

Ton Kalker

Philips Research & University of Eindhoven

Ton.Kalker@ieee.org

Overview

- Definition
- Why watermarking?
- Example
- Spread-Spectrum
- Matched Filtering
- Watermark parameters
- Attacks

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Watermarking =

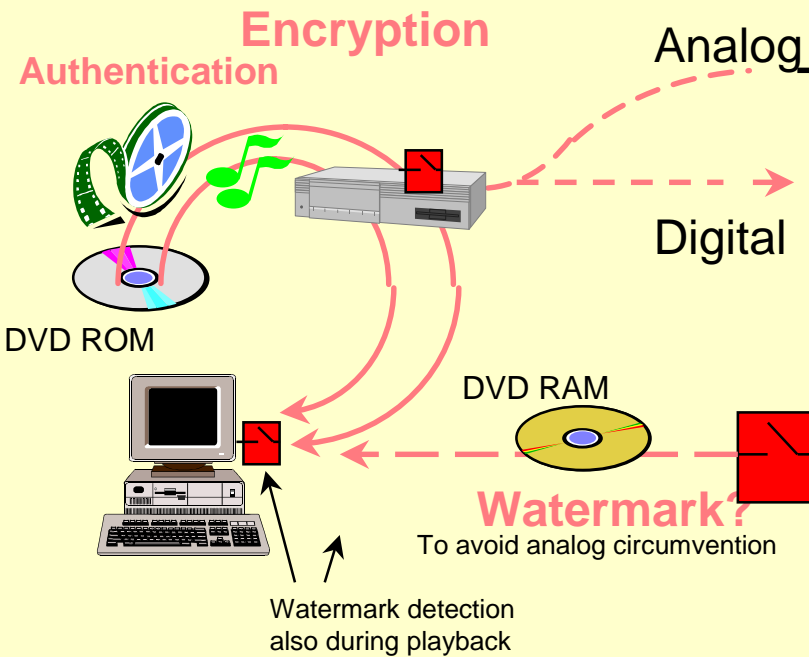
- The art of actively modifying audio-visual content such that the modifications
 - Are imperceptible (who is the listener?),
 - Carry retrievable information,
 - That survives under degradations of the content,
 - *And is difficult to remove & change by unauthorized users (cryptography).*
- Watermarking is not adding meta-data to header fields!

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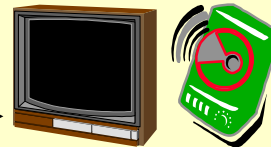
Compliant World

- All content is encrypted on all digital interfaces
- Link-by-link encryption; devices internally process clear content
- Controlled by CSS, 5C, 4C, ...
- Includes DVD players, DVD RAM, SDMI audio, DVD audio, PC's



Non-Compliant World

- All analog devices, some digital
- Marginalized by standardization efforts



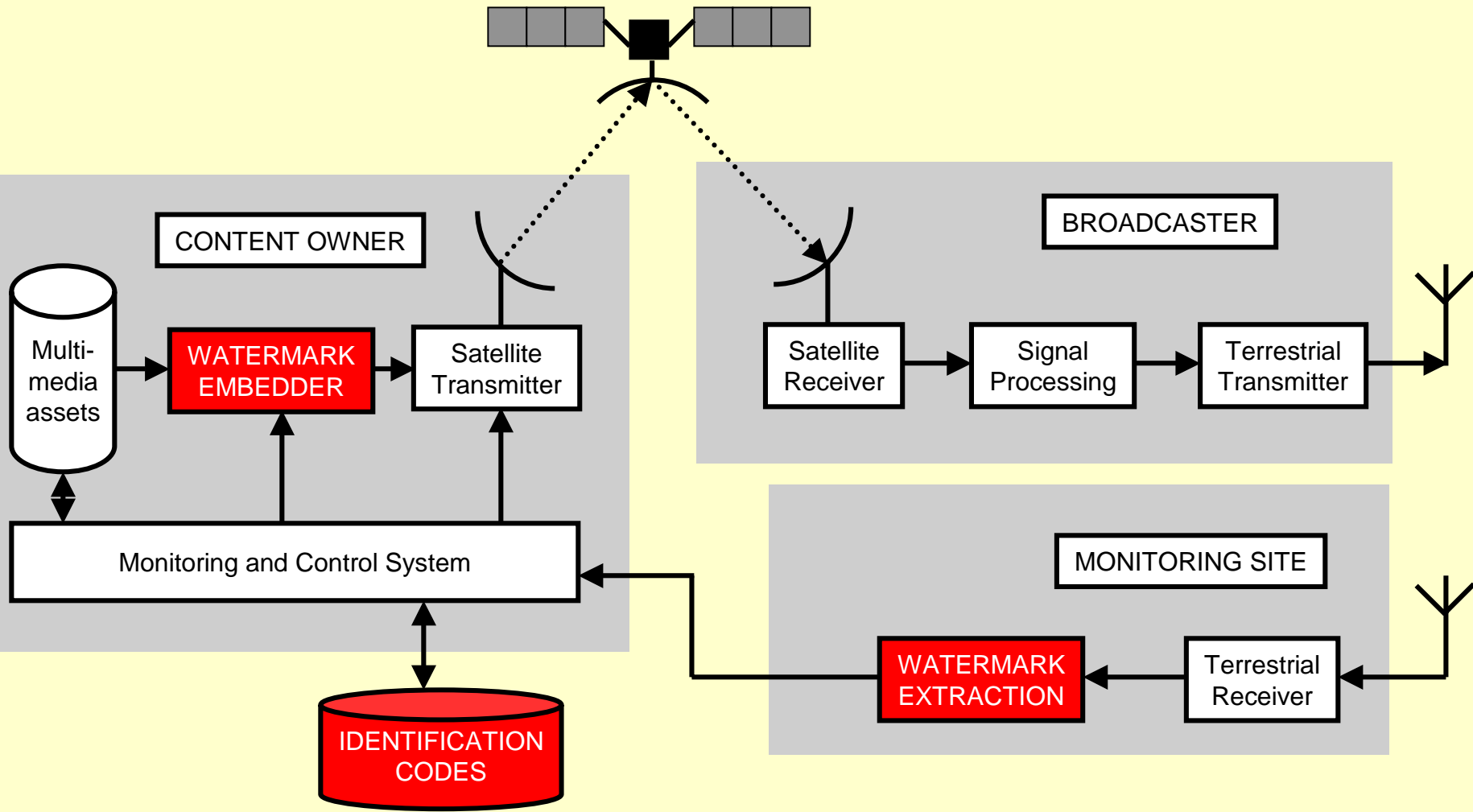
- Macrovision spoilers
- Watermarks

- By licensing contract no unprotected output

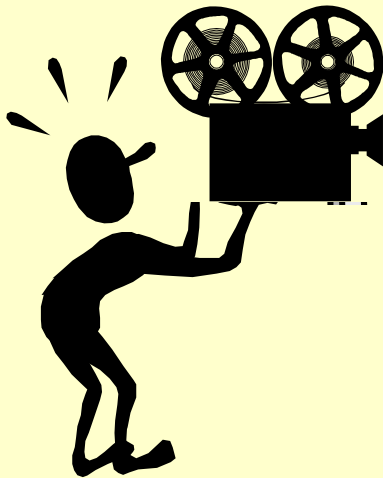


- New laws in US and EU

Broadcast Monitoring



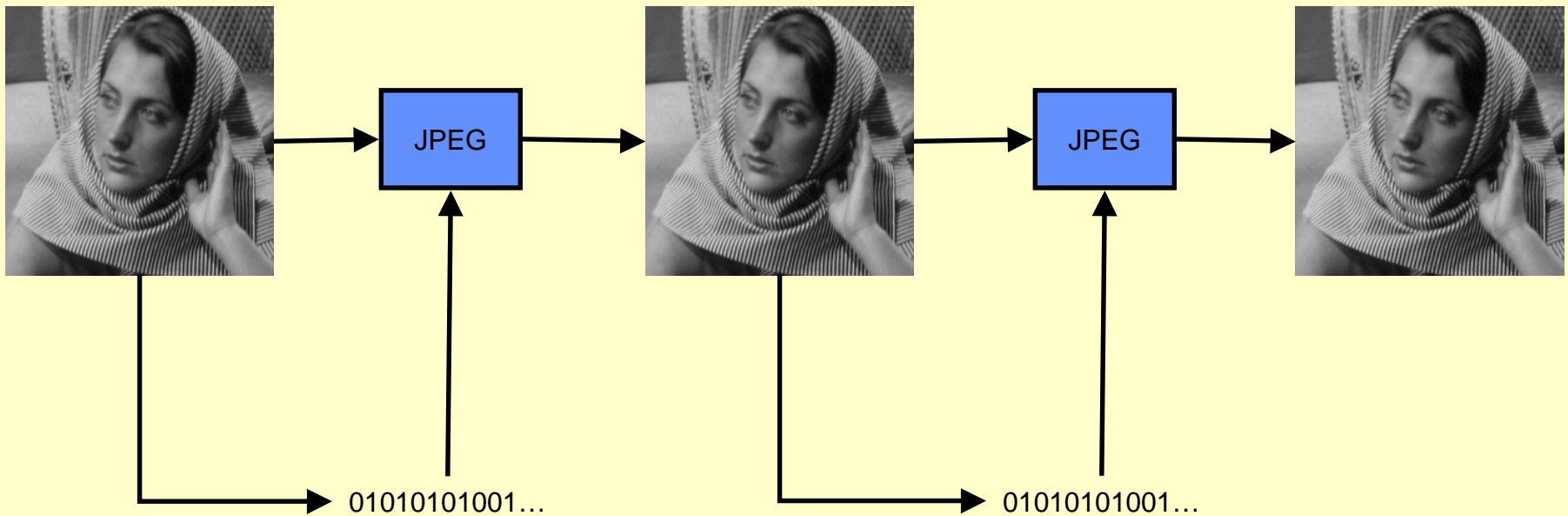
Digital Cinema



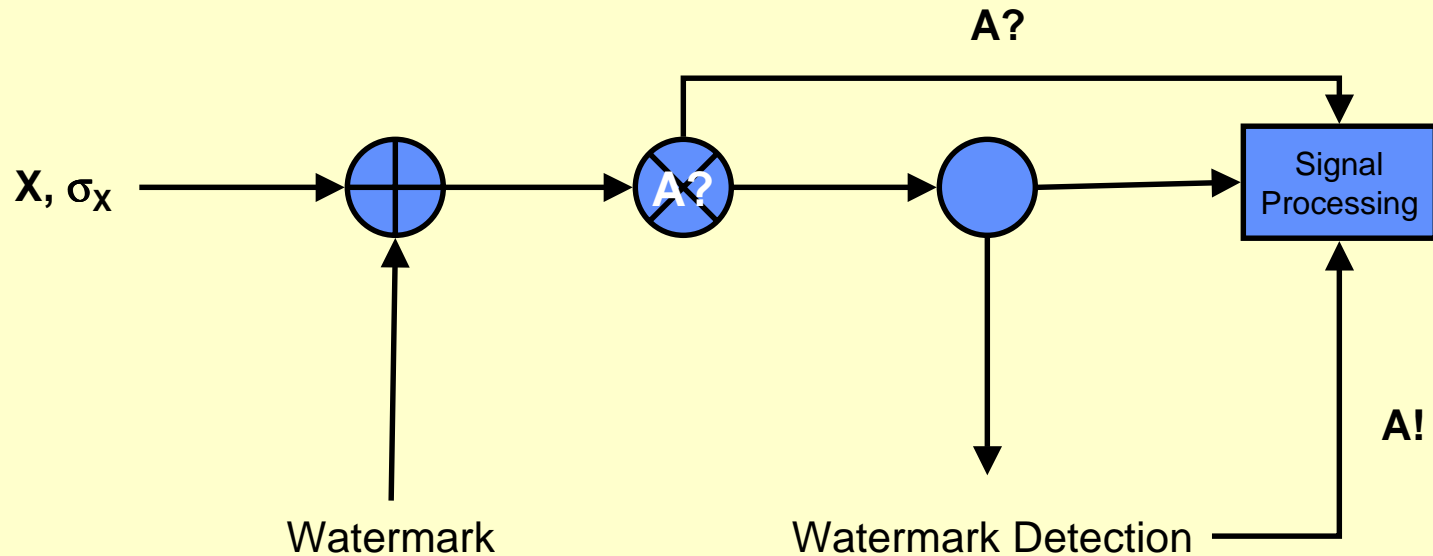
Name That Tune



Helper Data for Processing



Helper Data for Calibration

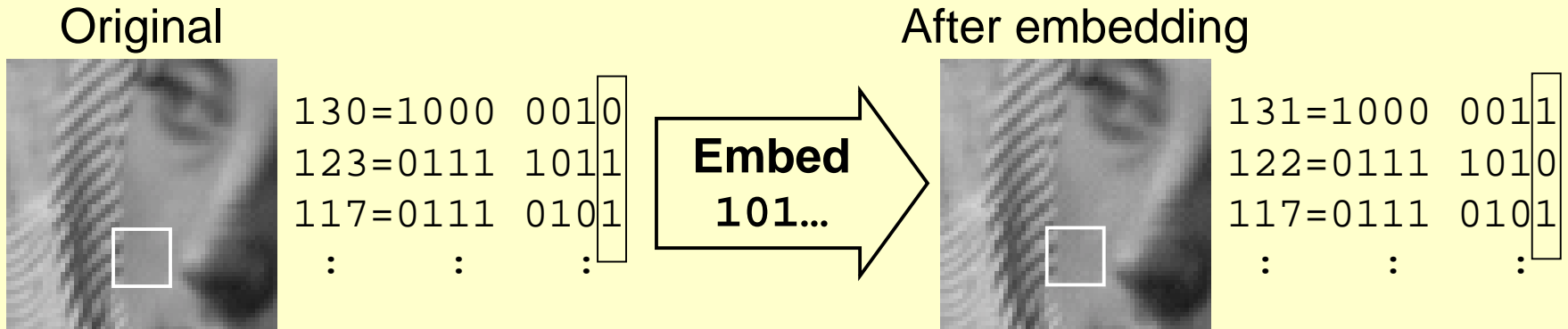


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- **Example**
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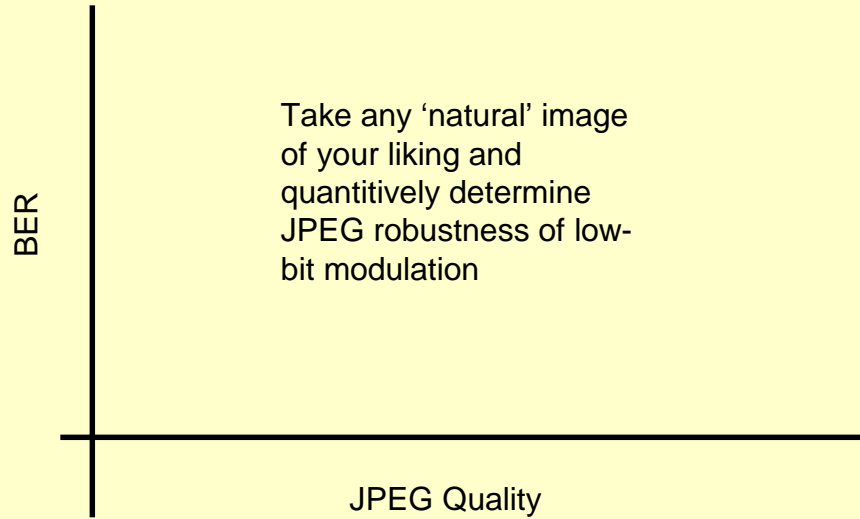
Low-bit Modulation

- Early scheme: alter LSB or low-order bits



- imperceptible (modify only LSBs)
- secure (encrypt embedded information)
- not robust (e.g., randomly set LSBs to 0 or 1)
- More accurate: secure info-hiding method

Low Bit Modulation



Patchwork

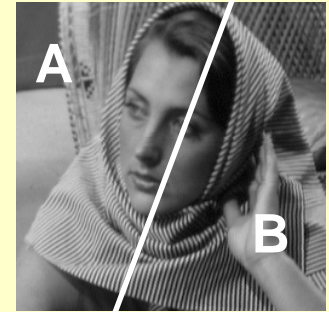
- 2 disjoint sets, A and B , of $N/2$ pixels each
 - pixels in each set (“patch”) chosen randomly
 - assumption:

$$S = \left(\sum_i A_i - \sum_i B_i \right) / N \approx 0$$

- embedding bit $b \in \{-1, +1\}$: $A'_i \leftarrow A_i + b * 1$, $B'_i \leftarrow B_i - b * 1$

$$\begin{aligned} S' &= \left(\sum_i A'_i - \sum_i B'_i \right) / N = \\ &= \left(\sum_i A_i - \sum_i B_i \right) / N + \\ &+ (N/2 - (-N/2)) / N \approx b \end{aligned}$$

- if $|S'| \approx 1$, watermark present with value $\text{sign}(S')$
- Prototypical spread-spectrum watermarking
 - communicate information via many small changes



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Spread-Spectrum Watermarking

- Original Signal $x[i]$ (Gaussian, iid, σ_X, \dots)
- Watermark $w[i]$ (Gaussian, iid, σ_W, \dots)
- Watermarked Signal
 - (1/2)-bit version (*copy protection*)
 - H0: $Y[i] = X[i]$
 - H1: $Y[i] = X[i] + W[i]$
 - 1-bit version (*helper data*)
 - H0: $Y[i] = X[i] - W[i]$
 - H1: $Y[i] = X[i] + W[i]$

Spread-Spectrum Watermarking

- Received Signal $Z[i]$
 - Distinguish between two hypotheses H_0 and H_1 .
- Maximum likelihood testing
 - (Gaussian, iid) optimal tests statistic given by correlation
 - $D = (\sum_i Z[i] W[i]) / N$
- Not Marked : $Z = X$
 - $E[D] = (\sum_i E[X[i]] E[W[i]]) / N = 0$
 - $E[D^2] = E[(\sum_i X[i] W[i])^2] / N^2 =$
 $= (\sum_i E[X[i]^2] E[W[i]^2]) / N^2 =$
 $= \sigma_X^2 \sigma_W^2 / N$

Spread-Spectrum Watermarking

- Marked : $Z = X + W$
 - $E[D] = \sigma_W^2$
 - $\sigma_D^2 = \sigma_X^2 \sigma_W^2 / N$
- For N large D is approximately Gaussian distributed
- Error rate determined by $Q(D / \sigma_D)$
- Marked : $E[D] / \sigma_D = \text{Sqrt}(N) (\sigma_W / \sigma_X)$
- Robustness increases with
 - More samples
 - More watermark energy
 - Less host interference

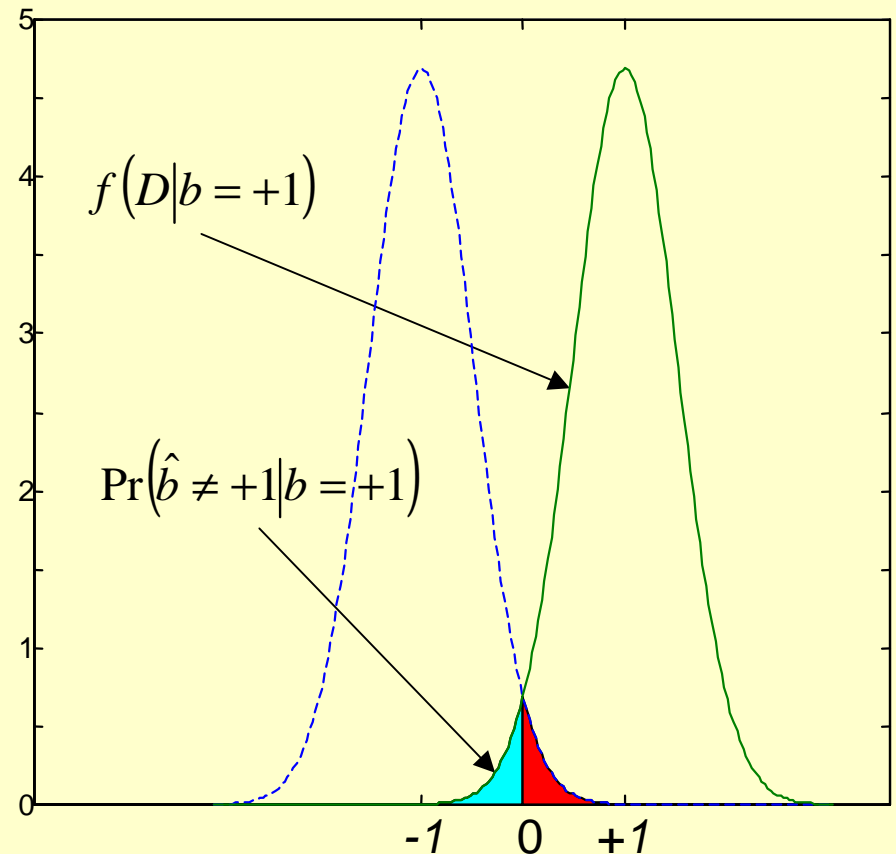
Detection (effectiveness)

- Correlation sum D
 - assumed Gaussian
 - $\sigma_w = 1$
 - variance $\sigma_x^2/(N)$
- Decision rule becomes

$$\hat{b} = \begin{cases} +1, & \text{if } D > 0; \\ -1 & \text{if } D < 0. \end{cases}$$

- Probability of error
 - Q function

$$Q\left(\frac{\sqrt{N}}{\sigma}\right)$$



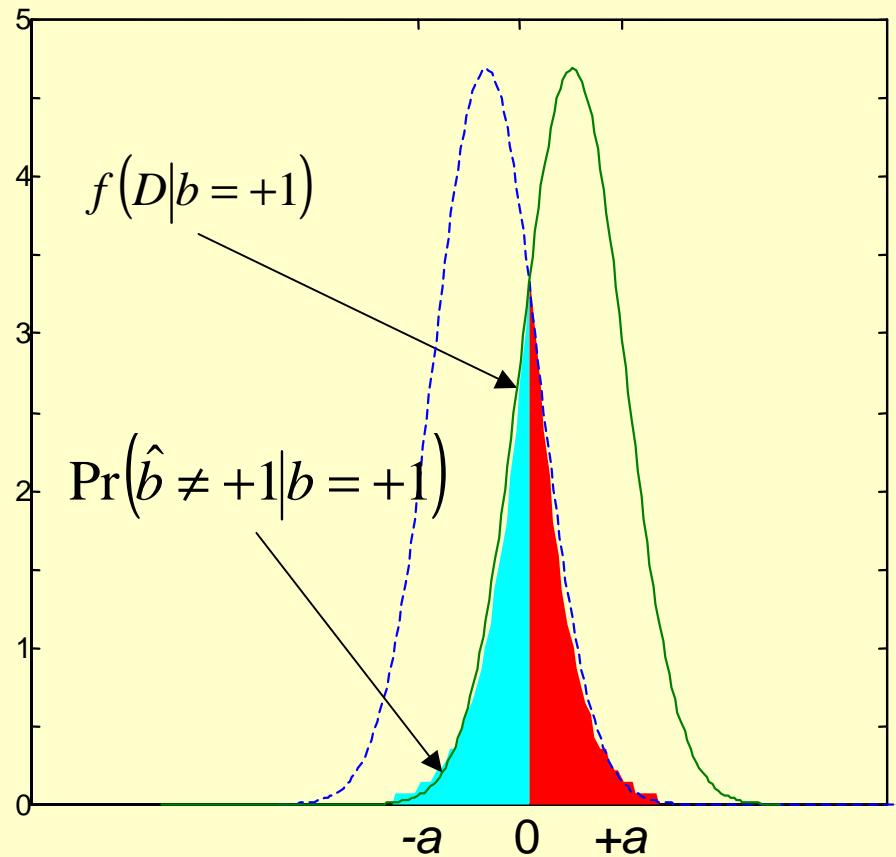
Detection (robustness)

- Correlation sum D
 - assumed Gaussian
 - mean $-a, +a$
 - variance $\sigma_x^2/(N)$
- Decision rule becomes

$$\hat{b} = \begin{cases} +1, & \text{if } D > 0; \\ -1 & \text{if } D < 0. \end{cases}$$

- Probability of error
 - Q function

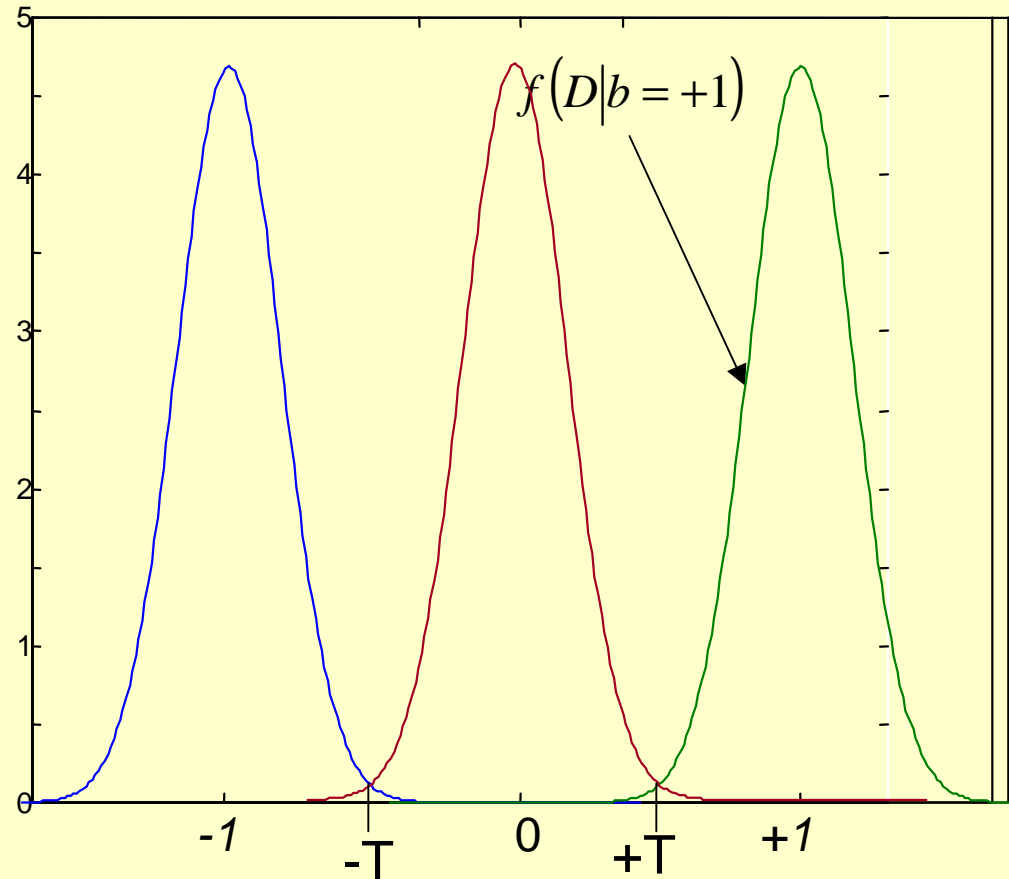
$$Q\left(a \frac{\sqrt{N}}{\sigma}\right)$$



Detection (false positives)

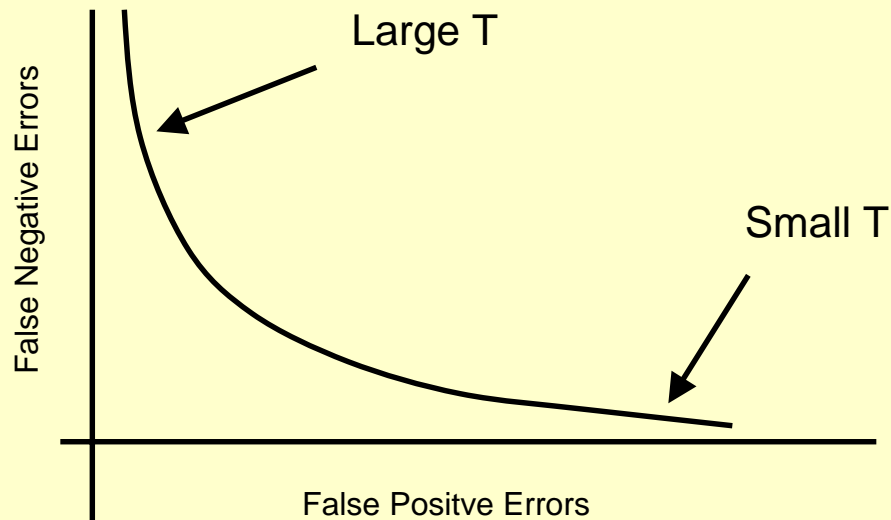
- Correlation sum D
 - assumed Gaussian
 - mean $-1, 0, +1$
 - variance $\sigma_x^2/(N)$
- Decision rule becomes
$$\hat{b} = \begin{cases} +1, & \text{if } D > +T; \\ -1, & \text{if } D < -T; \\ 0, & \text{if } |D| \leq T. \end{cases}$$
- Probability of false positive

$$2Q\left(T \frac{\sqrt{N}}{\sigma}\right)$$

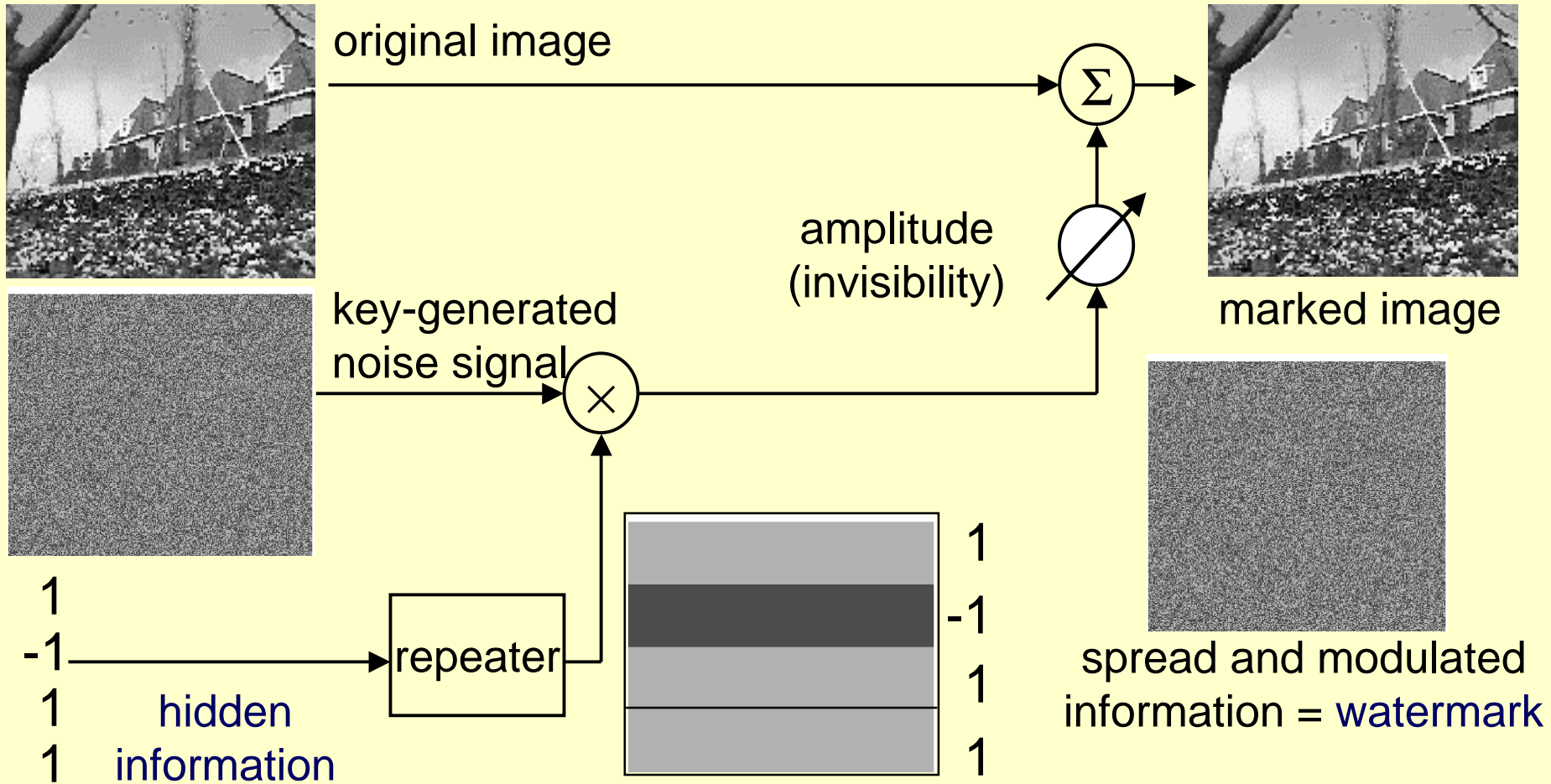


Spread-Spectrum Watermarking

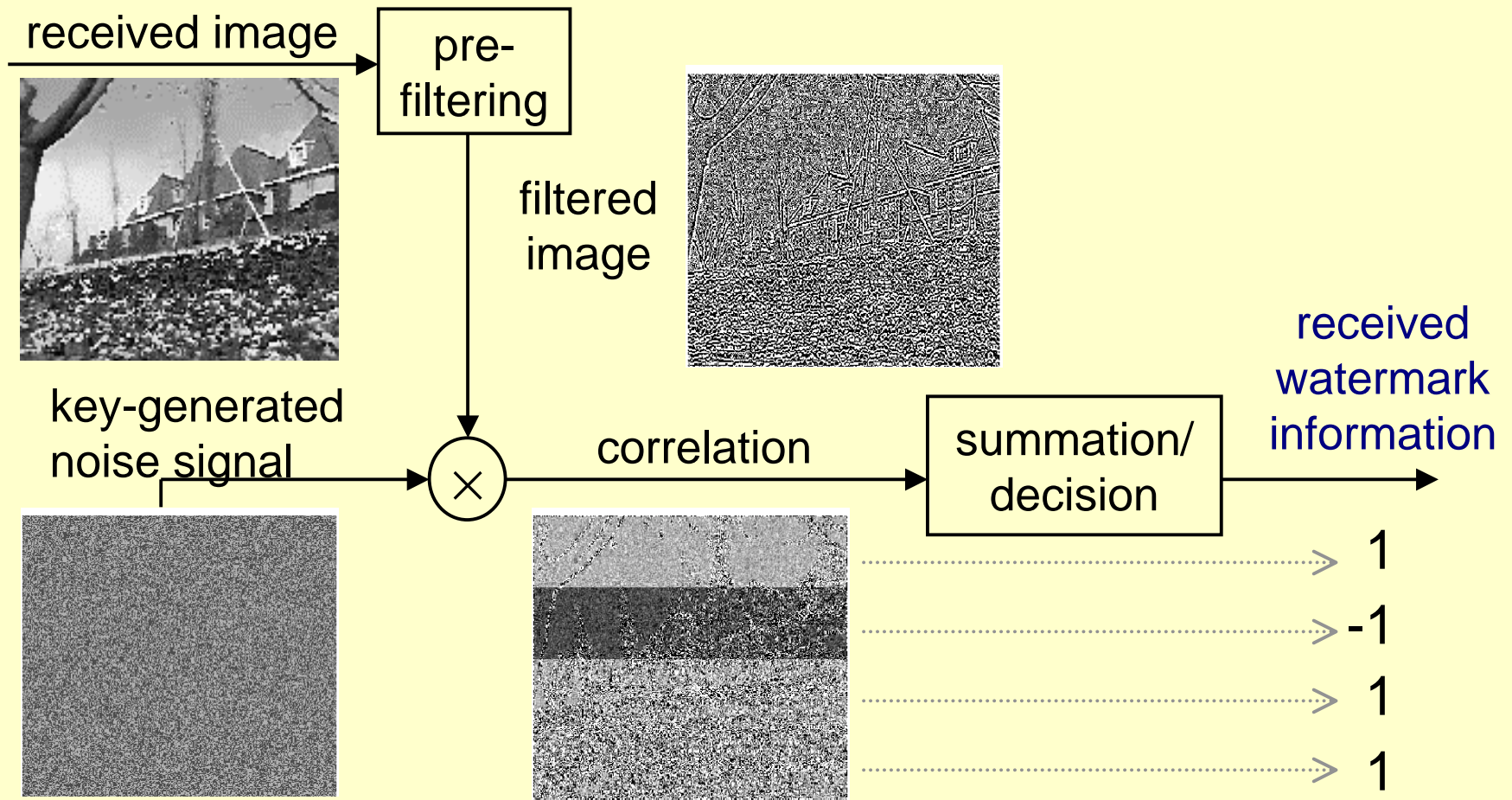
Receiver-Operator Characteristic (ROC) Curve



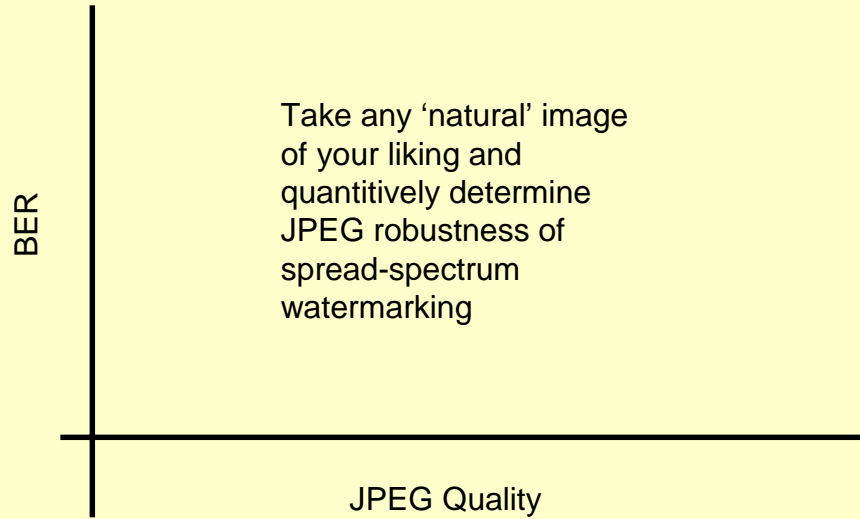
Watermark Embedding



Watermark Retrieval



Spread-Spectrum Watermarking

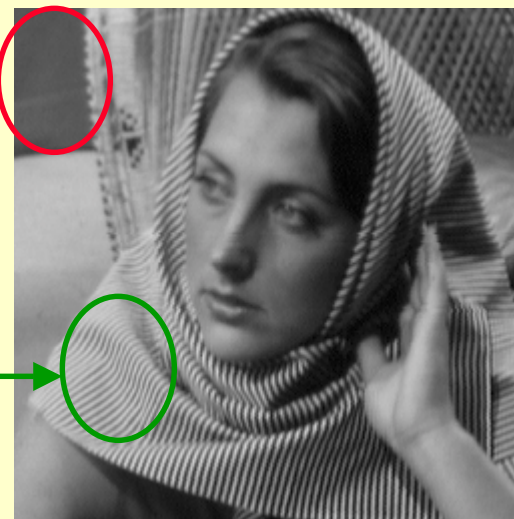


Perceptual Watermarking

- Original x .
- Apply transform T : $y = T(x)$
 - $T = I, \text{DCT}, \text{FFT}, \text{log}, \dots$ (or any combination thereof)
- Add pseudo-random sequence w : $z = y + w$
 - Allow adaption of w to host signal
 - $Z = Y + \alpha W$
 - In position
 - only in textured image regions, not in silence
 - In value
 - less energy in flat regions than in textured regions
- Apply inverse transform: $m = T^{-1}(z)$

Perceptual Watermarking

- Example: PatchWork
- $T = I$
 - Spatial watermarking
- $W = X_A - X_B$
 - Binary $\{-1,+1\}$ -valued pseudo-random sequence
- Adaptation, e.g.
 - Less power in flat regions
 - More power in textured regions



Perceptual Watermarking

- Received data m'
- Apply inverse transform $T^{-1}: z' = T^{-1}(m')$
- Assume $z' = y' + h*w$
 - Hypothesis testing
 - $h = 0$: not watermarked
 - $h = 1$: watermarked
- Determine optimal detector
 - Prefilter + correlation
 - $D = \langle y', w \rangle + h \langle w, w \rangle$

Popular Example: NEC Scheme

- Heuristic claim

- watermark should be embedded in the “perceptually significant frequency components” for best robustness

- Embedding

- N watermark samples $w_i \sim N(0,1)$; e.g., $N = 1000$
- embed in the N largest-amplitude DCT coefficients (except DC coefficient) x_i

$$y_i = x_i(1 + \alpha w_i)$$

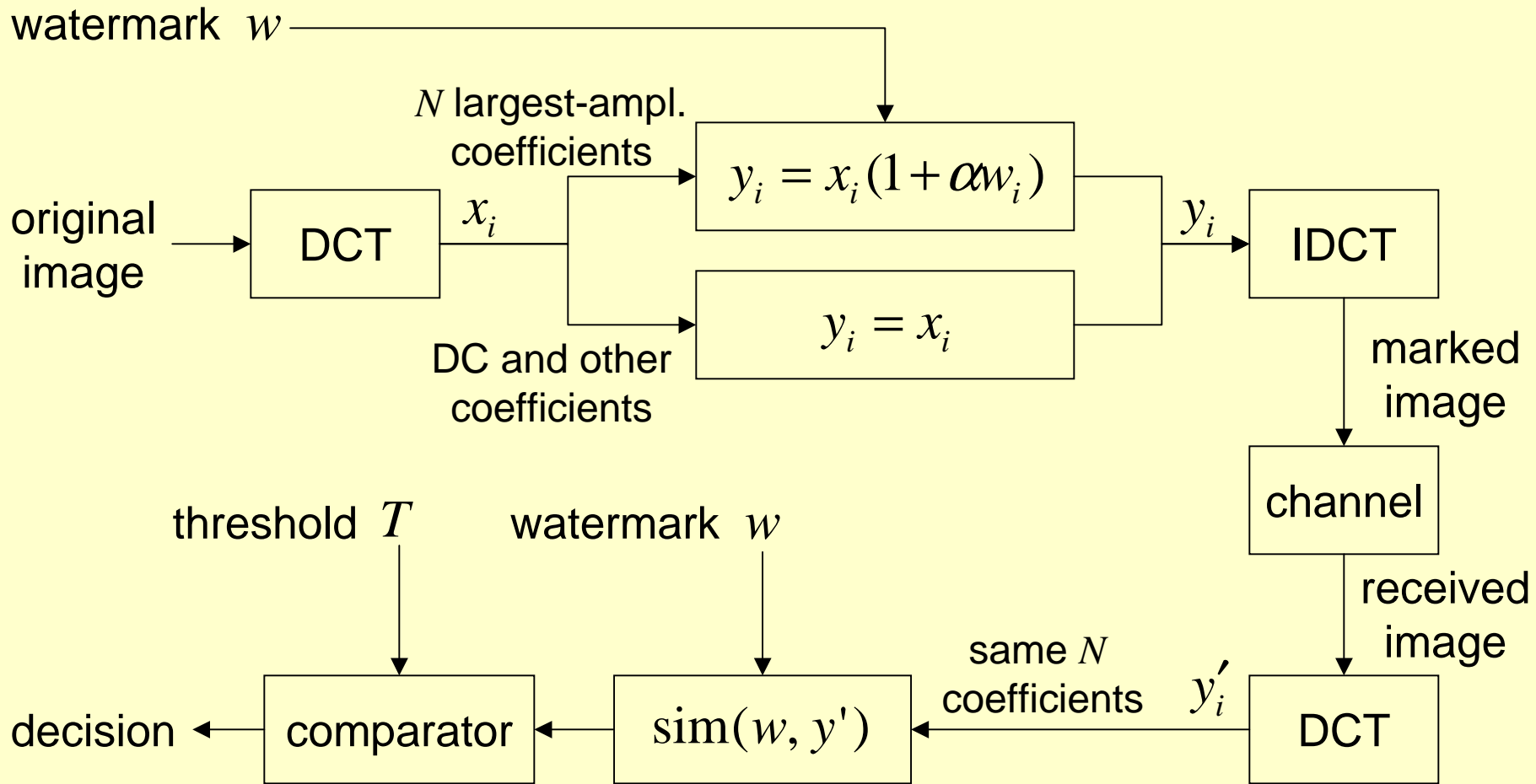
- Detection

- extract the same N DCT coefficients y'_i
- compute the similarity (normalized correlation) between y'_i and w_i

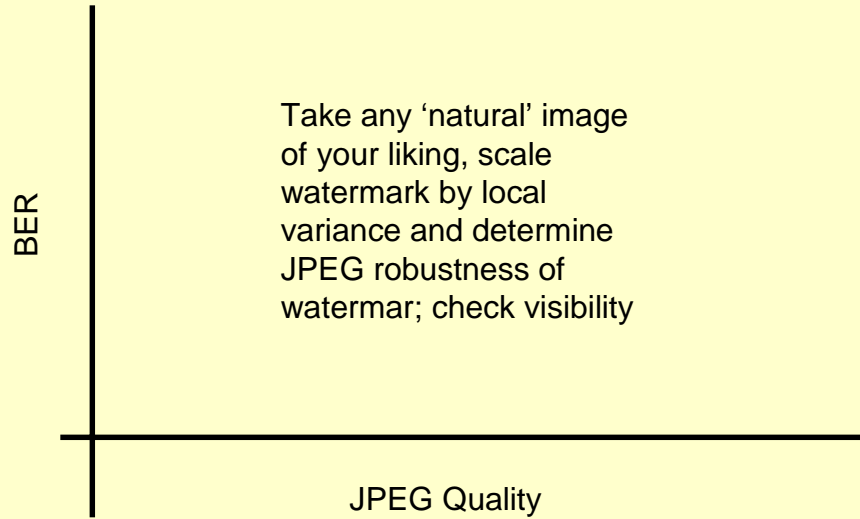
$$\text{sim}(w, y') = \frac{\langle w, y' \rangle}{\sqrt{\langle y', y' \rangle}}$$

- watermark w is present if $\text{sim}(y', w) > T$

Block Diagram of NEC Scheme



Perceptual Watermarking



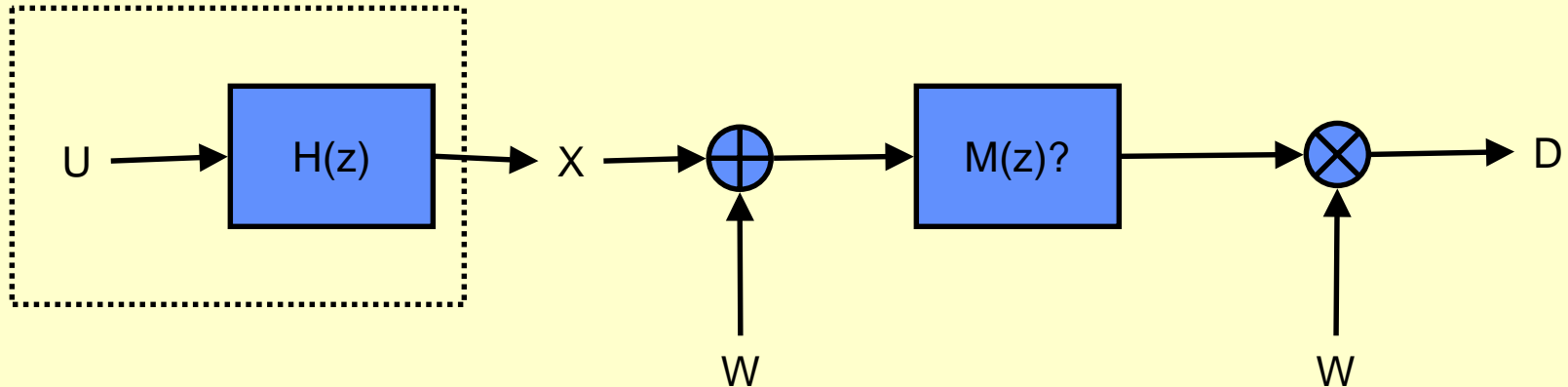
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Matched Filtering

- Audio-visual data are usually not well modelled as Gaussian iid sources!
- For images (for neighbouring pixels)
 - $E[X[i] X[i+1]] / \sigma_X^2 \approx 0.9$
- Better model $X = H * U$, where
 - H is low pass
 - U is random iid source
- Example : $X[i+1] = a X[i] + U[i+1]$
 - $a \approx 0.9$
 - $H(z) = (1 - a z^{-1})^{-1}$

Matched Filtering



- Correlation in z-domain notation
 - $A(z) = \sum a_i z^{-i}$
 - $[A(z)]_0 = a_0$
 - $\sum a_i b_i = [A(z) B(z^{-1})]_0$
- $D = [(M(z) H(z) U(z) + M(z) W(z)) W(z^{-1})]_0$

Matched Filtering

- Cost function

- $C_M =$

- $= (\text{Righthand term})^2 / E[\text{variance lefthand term}]$

- $= [M(z)W(z) W(z^{-1})]_0^2 / E[[M(z) H(z) U(z) W(z^{-1})]_0^2]$

- Simplification

- $C_M =$

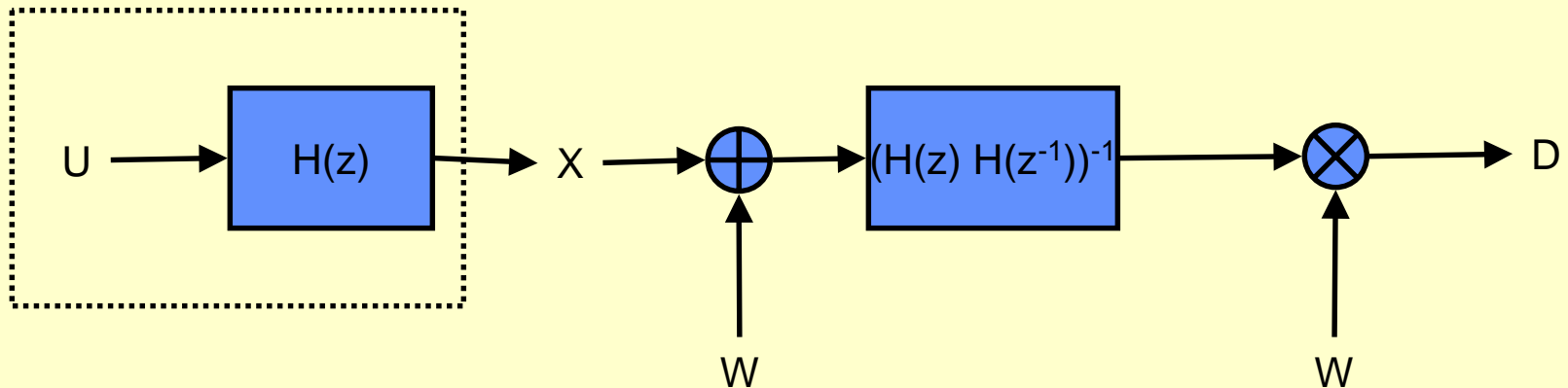
- $= (N^2 [M(z)]_0 \sigma_W^4) / (N \sigma_W^2 \sigma_U^2 [M(z) M(z^{-1}) H(z) H(z^{-1})]_0)$

- $= N (\sigma_W^2 / \sigma_U^2) ([M(z)]_0 / [M(z) M(z^{-1}) H(z) H(z^{-1})]_0)$

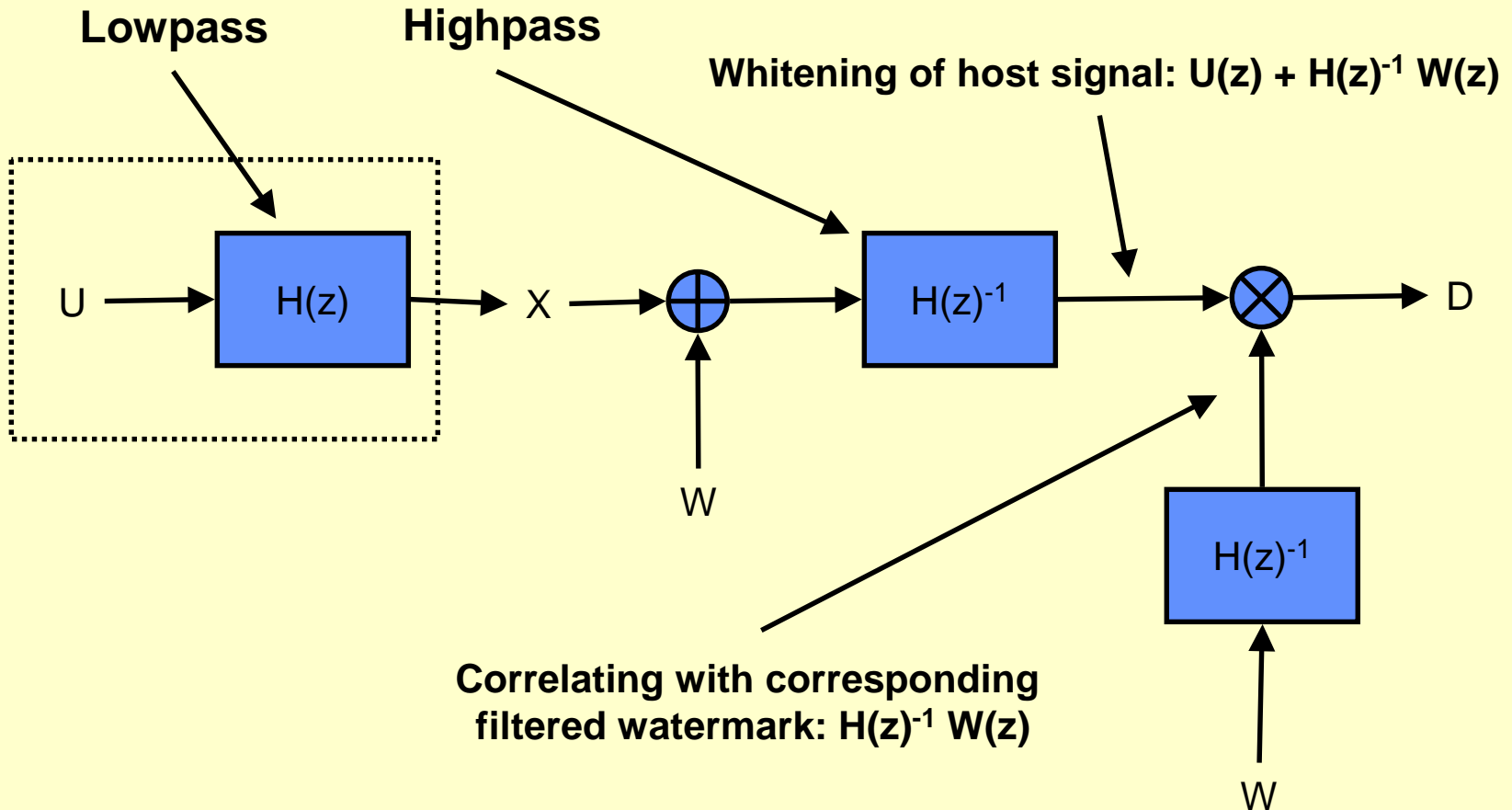
Matched Filtering

- Optimize in the frequency domain
 - $\mu_i = M(\omega_i)$, $\eta_i = H(\omega_i)$
 - $C_M = \Sigma \mu_i / (\Sigma \mu_i^2 \eta_i^2)$
 - We may assume $\Sigma \mu_i = 1$
 - Using Lagrange multipliers we find
 - $\mu_i = 1 / \eta_i^2$
 - $M(z) = (H(z) H(z^{-1}))^{-1}$

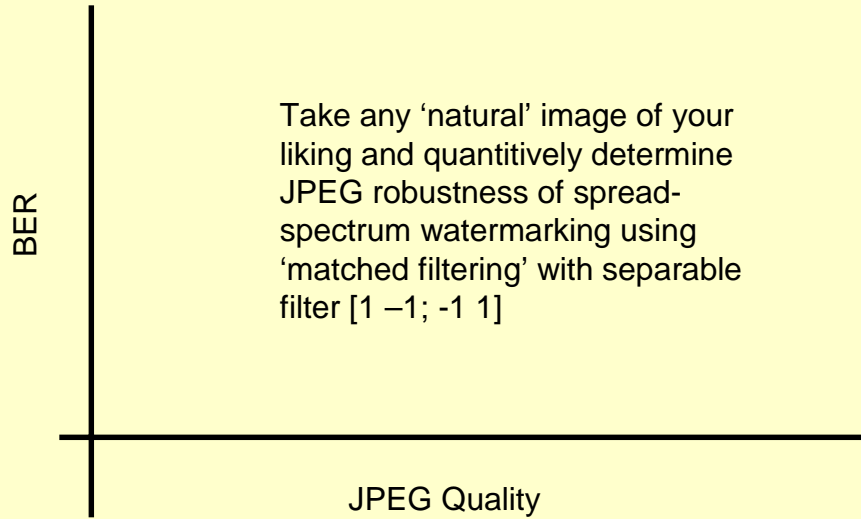
Matched Filtering



Matched Filtering



Spread-Spectrum Watermarking



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Watermark Parameters

- Perceptibility
 - perceptibility of the watermark in the intended application



Original image



Image + hidden information

Watermark Parameters

- Robustness
 - resistance to (non-malevolent) quality respecting processing



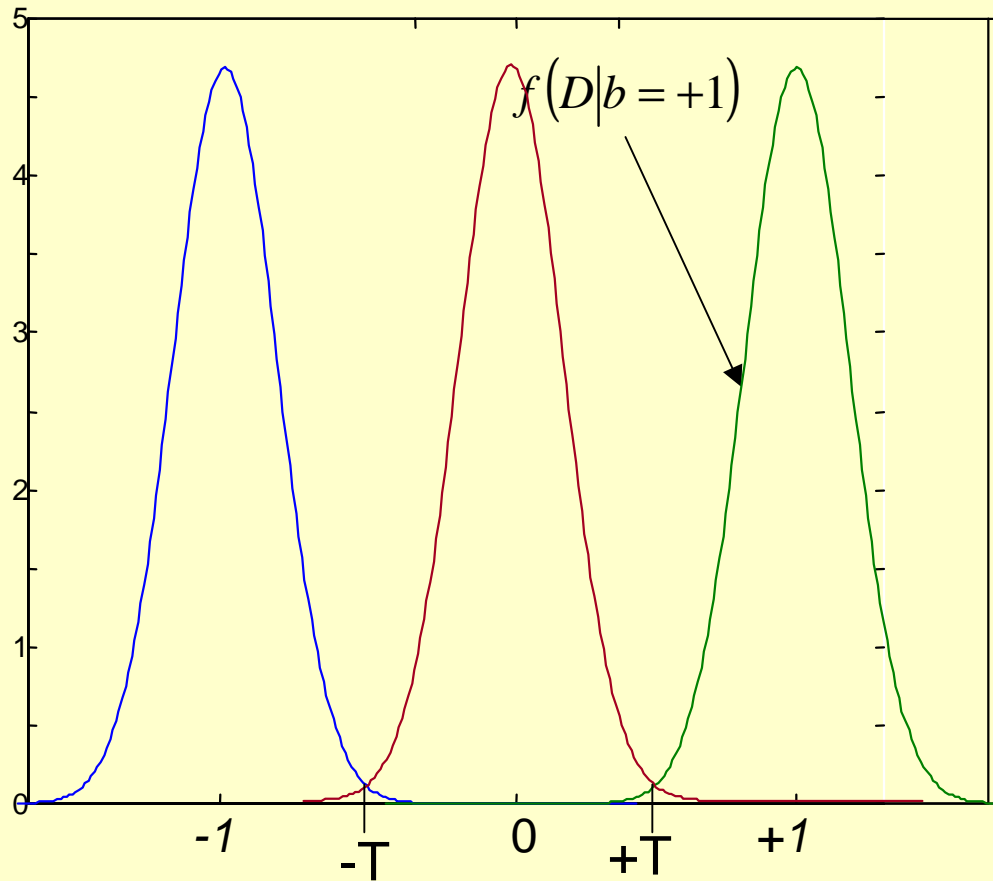
JPEG compression



Additive noise & clipping

Watermark Parameters

- Error Rates



Watermark Parameters

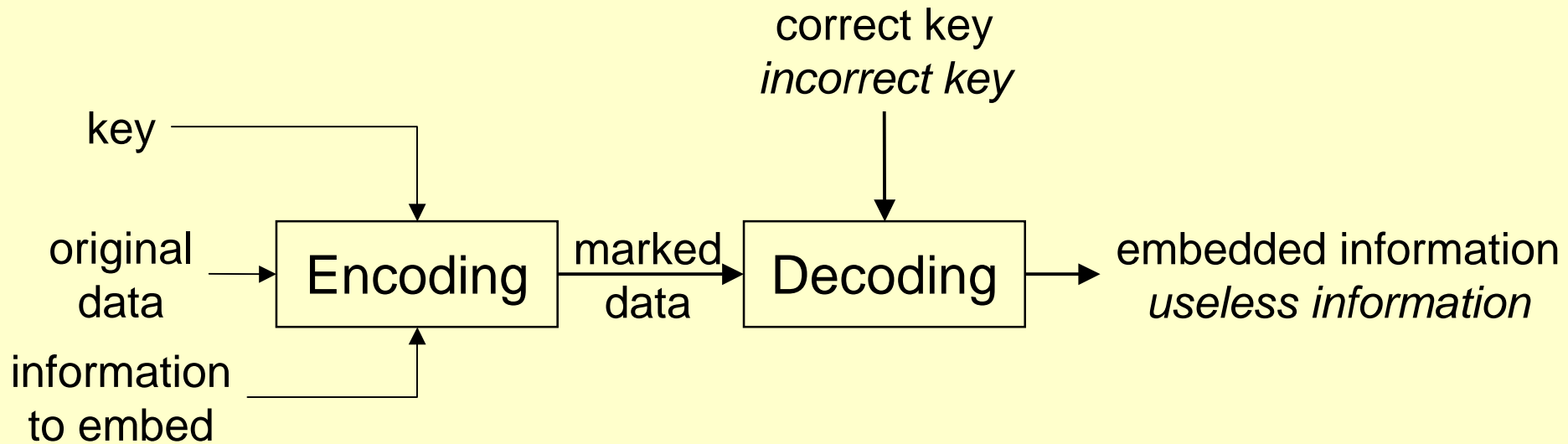
- Complexity
 - hardware & software resources, real-time aspects
 - baseband vs. compressed domain
- Granularity
 - minimal spatio-temporal interval for reliable embedding and detection
- Capacity
 - related to payload
 - #bits / sample

Watermark Parameters

- Layering & remarking
 - watermark modification
- Security
 - vulnerability to intentional attacks
 - Kerckhoffs' principle

Security

- Embedded information cannot be **detected**, **read** (**interpreted**), and/or **modified**, or **deleted** by unauthorized parties
- Kerckhoff's principle: Security resides in the secrecy of the key, **not** in the secrecy of the algorithm.

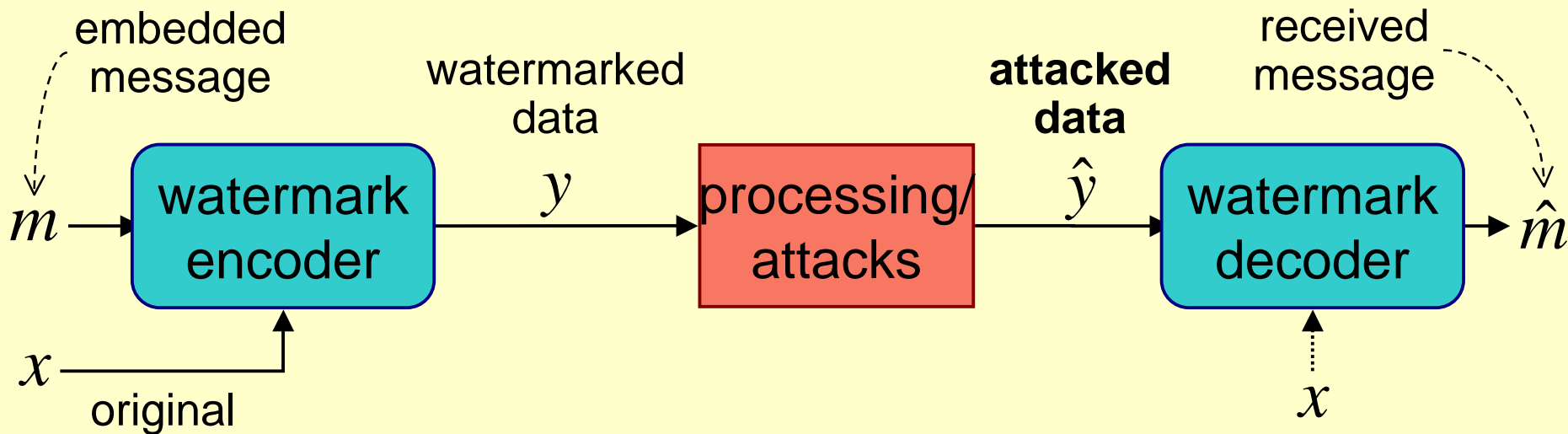


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Attacks and Communications Viewpoint

- Watermarked data will likely be processed
- **Attack** - any processing that may coincidentally or intentionally damage the embedded information
- Treat attacks like a communications channel



Evaluating Robustness

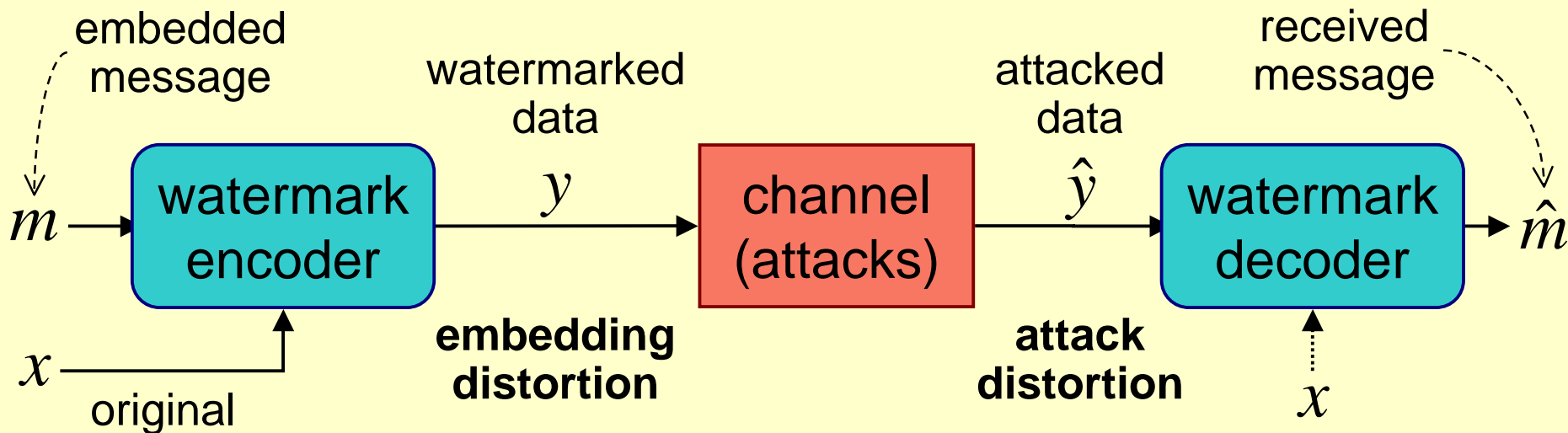
- Robustness: easy to define, hard to evaluate
 - Embedded information cannot be damaged or destroyed without making the attacked data useless
 - How to evaluate robustness in a well-defined sense?

*“A watermark is robust if **communication** cannot be impaired without rendering the attacked data **useless**.”*

- Kerckhoff's principle
 - Assume opponent has complete knowledge of your strategy (algorithm and implementation) but lacks a secret (key).

Need for a Distortion Measure

- When is the attacked data useless?
- Quantify “usefulness” of attacked data
- Multimedia → measure distortion of attacked data
 - inherently subjective, always debatable
 - imperfect but measurable

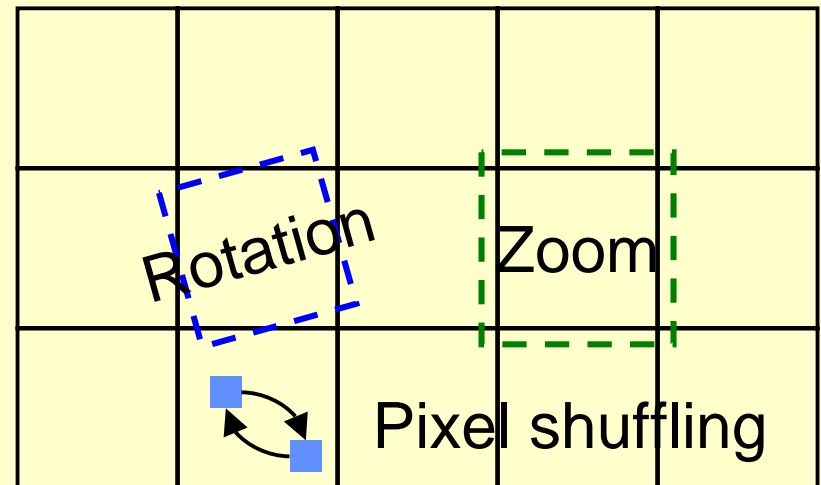


Classes of Attacks

- Simple waveform processing
 - “brute-force” approach
 - impairs watermark and original data
 - compression, linear filtering, additive noise, quantization
- Detection-disabling methods
 - disrupt synchronization
 - **geometric transformations** (RST), cropping, shear, re-sampling, shuffling
 - watermark harder to locate
 - distortion metric not well defined
 - meaning of watermark presence?
 - change of ROC curve!
- Advanced jamming/removal
 - intentional processing to impair/defeat watermark
 - **watermark estimation**, collusion (multiple copies)
- Ambiguity/deadlock issues
 - reduce confidence in watermark integrity
 - creation of fake watermark or original, estimation and copying of watermark signal

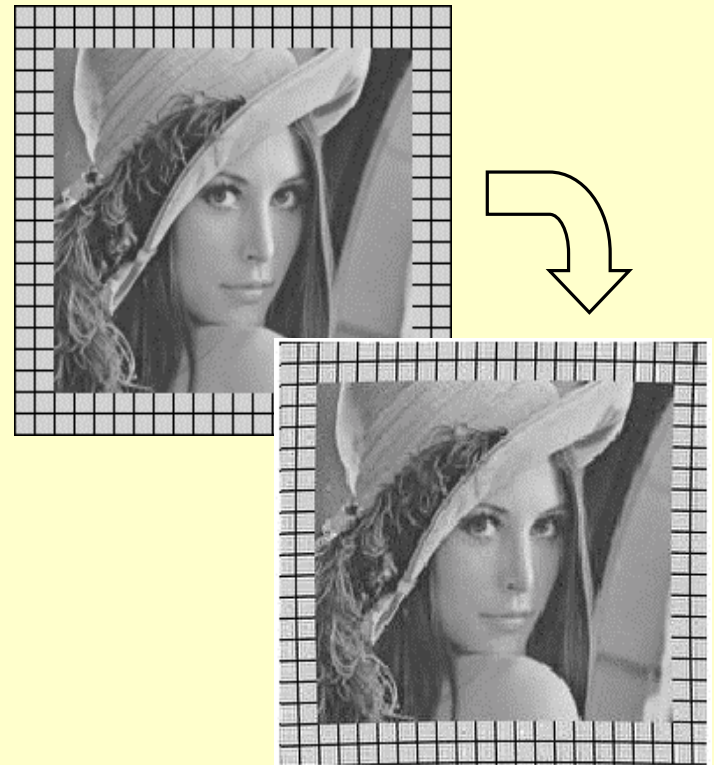
De-synchronization

- Attack
 - harder to find watermark
 - does not remove watermark
- How to measure distortion?
- Spread spectrum
 - fails without sync
 - re-synchronizing difficult
 - noiselike carrier
 - no peaks in frequency



StirMark

- Popular, free WWW software
 - simulate printing and scanning
 - nonlinear geometric distortion + JPEG
- Easy to use and test
- Limitations
 - features available elsewhere
 - purely empirical
 - does not suggest how to improve system
 - does not use Kerckhoff's principle!
 - does not target system weaknesses
 - suboptimal attack
 - false sense of security



Resynchronization Methods

- Use of templates
 - pattern of peaks in frequency domain
 - attacker can locate pattern, too!
 - pattern of local extrema
 - harder for attacker to locate or recognize
 - harder for receiver, too
 - seeking pattern is like seeking watermark signal
- Invariant representations
 - translation invariance
 - Fourier magnitude
 - rotation and scale invariance
 - log-polar mapping
$$(x, y) \leftrightarrow (\mu, \theta)$$
$$x = e^{\mu} \cos \theta, y = e^{\mu} \sin \theta$$
 - Fourier-Mellin transform
 - cannot handle aspect ratio changes, shear, etc.

Translation Robustness

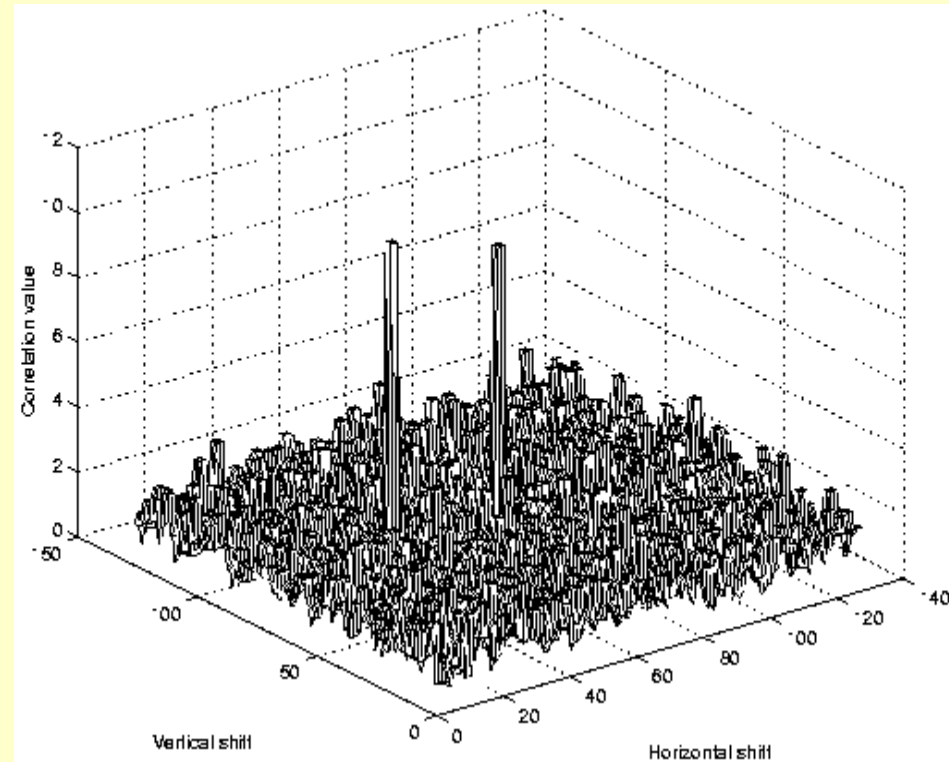
- Original Marked Data
 - $Y[i] = X[i] + W[i]$
- Translated Data
 - $Z[i] = Y[i+k]$
- Detector strategy (k unknown)
 - Trial and error: correlating at shifted positions
 - $D[i] = \sum_l Z[i-l] W[i]$ (exhaustive search)
 - $D(z) = W(z) Z(z^{-1})$
 - Efficient computation with Fourier transform!
 - $D = \text{FFT}^{-1}(\text{FFT}(W) * \text{FFT}^*(Z))$

Translation Robustness

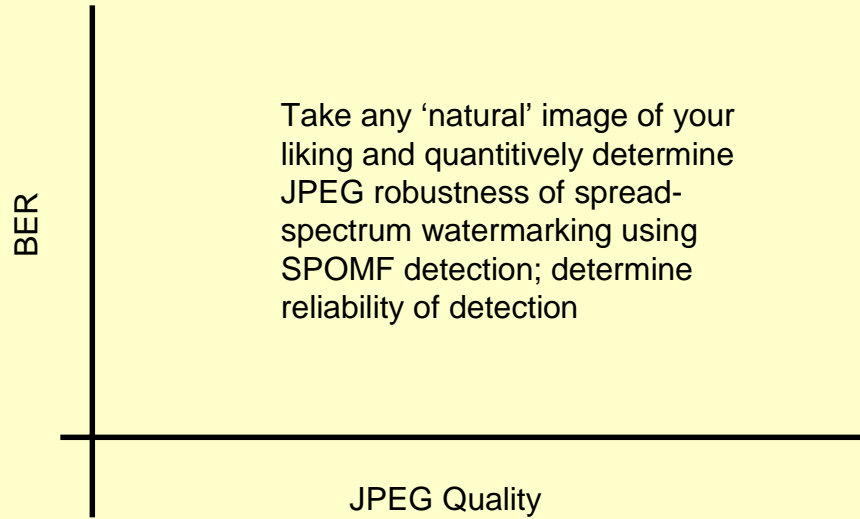
- Integration with Matched Filter
 - $D = \text{FFT}^{-1}(\text{FFT}(W) * \text{FFT}^*(Z) * |\text{FFT}(H)|^{-2})$
 - In many cases, W and H are fixed and their Fourier transforms can be pre-computed and stored.
- Experimentally, retaining only phase information
 - Symmetrical Phase-Only Matched Filtering (SPOMF)
 - $D = \text{FFT}^{-1}(\text{Phase}(\text{FFT}(W)) * \text{Phase}(\text{FFT}^*(Z)))$
 - $\text{Phase}(a e^{2\pi i \omega}) = e^{2\pi i \omega}$

Translation Robustness

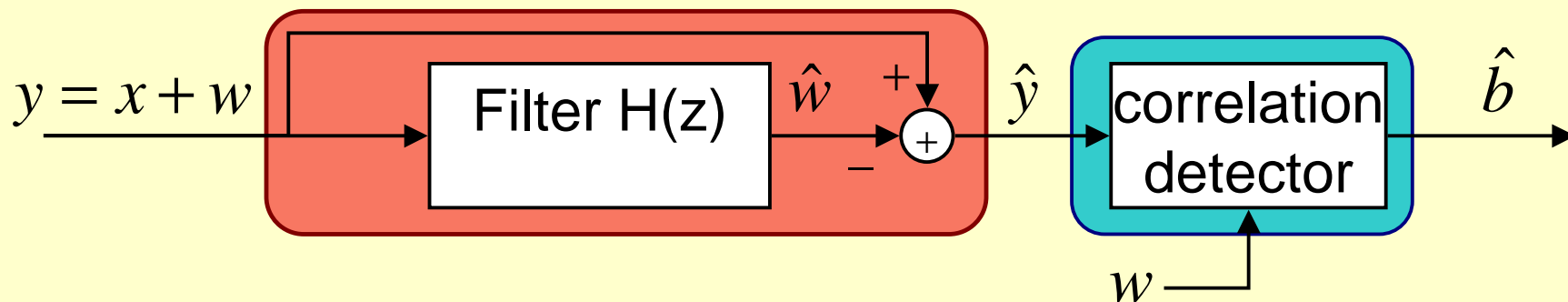
- Most values $D[i]$ correspond to non-synchronized watermark detections!
- $D(z)$ provides an estimate of the reliability of the watermark detection
- Reliability = $|\text{peak value(s)}| / \sigma_{\text{noise}}$



Translation Robustness



Estimation and Removal



- Problem Statement: find watermark $W[i]$ such that for given embedding distortion $N\sigma_w^2$ the detection reliability D and attack distortion D_a are maximized for any estimation filter $H(z)$.

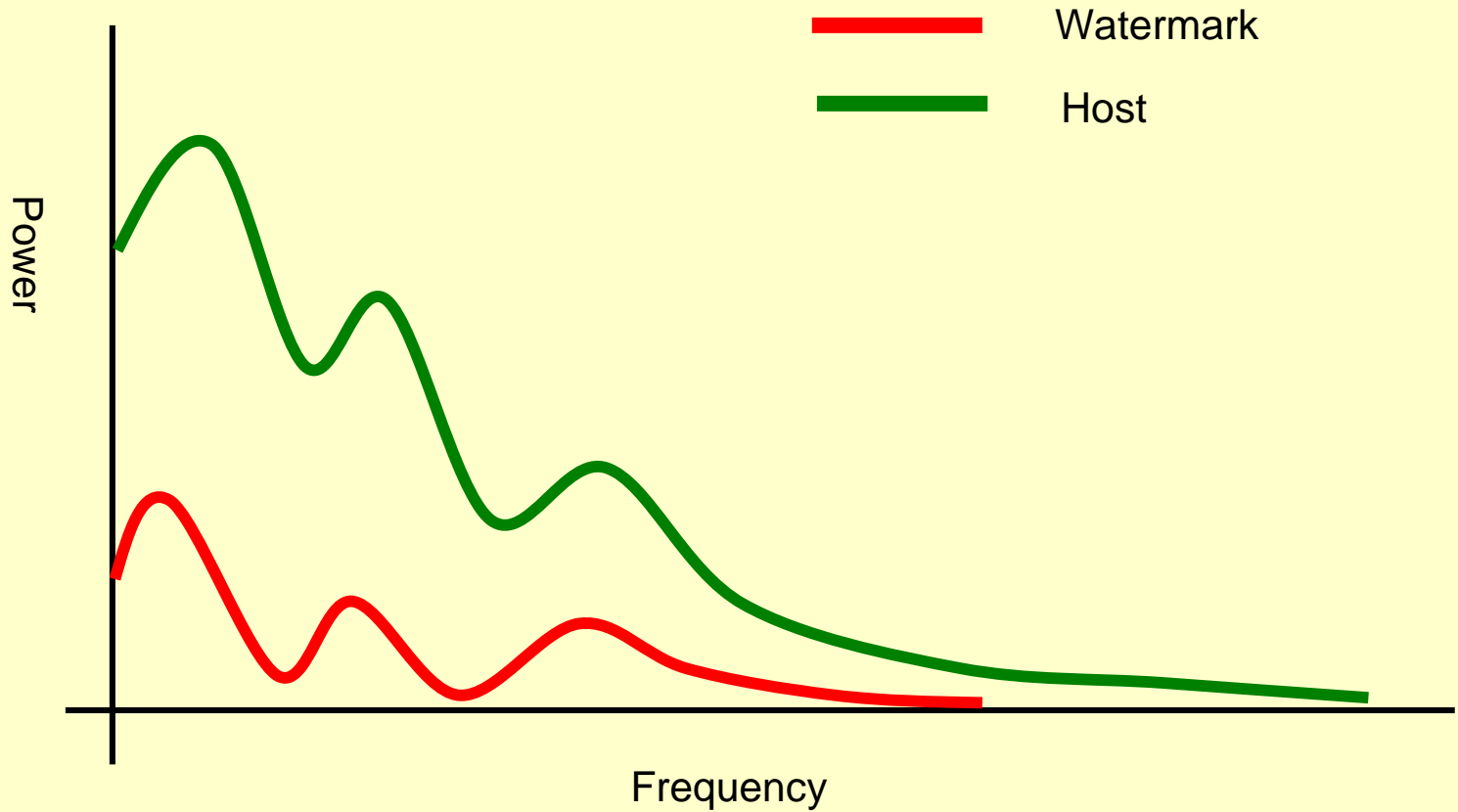
Estimation and Removal

- Problem description in frequency domain
 - $H(z) : \eta_i$
 - $W(z) : \omega_i$
 - $X(z) : \xi_i$
- Conditions
 - $\sum \omega_i^2 = N\sigma_W^2$
- Maximize
 - Attack distortion: $\sum (1 - \eta_i)^2 \xi_i^2 + \eta_i^2 \omega_i^2$
 - Detection reliability: $(\sum \eta_i \omega_i^2)^2 / \sum \eta_i^2 \xi_i^2$

Estimation and Removal

- From detection reliability (using Lagrangian multipliers)
 - $\eta_i = a \omega_i^2 / \xi_i^2$
- From attack distortion and condition (using Lagrangian multipliers)
 - $\eta_i = \eta = b \omega_i^2 / (\xi_i^2 + \omega_i^2)$
- Combining we find for all frequency components
 - $\rho = \omega_i^2 / \xi_i^2$ is fixed
- **Power Spectral Condition (PSC)** of [Su, Girod]
- Theoretical justification for heuristic arguments
 - Cox et al.

Power Spectral Condition



Estimation and Removal

- Optimal Watermark and Attack Filter
 - $\Phi_W = (\sigma_W^2 / \sigma_X^2) \Phi_X$
 - $H = \Phi_W / (\Phi_W + \Phi_X) = \sigma_W^2 / (\sigma_X^2 + \sigma_W^2)$ (scalar!)
- First example of game theory in watermarking
 - Embedder wants to maximize robustness
 - Tool: $W(z)$, Cost: Embedder distortion
 - Attacker wants to minimize robustness
 - Tool: $H(z)$, Cost: Attacker distortion

Estimation and Removal

