Image Steganography in Cryptography

組別:flag_t34m_n4me_1s_3ncrypt3d

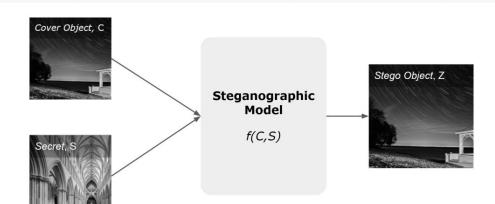
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Contents

- Recap: Image Steganography
- LSB
- DCT
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- Deep Learning
- Evaluation
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Recap: Image Steganography

- Hiding data behind digital images
- Common example: watermark



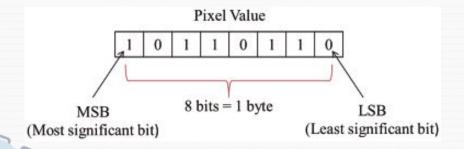




source: LSB 隱寫術:在圖片裡偷渡敏感資訊的方法

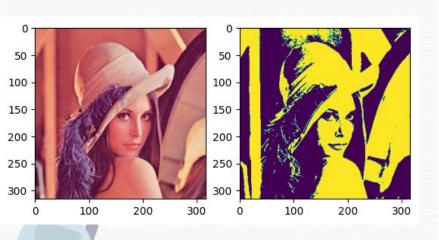
LSB (Least Significant Bits)

- The subtle differences in the image are difficult to detect with the naked eye.
- Color model: RGB, HSV...





LSB (Least Significant Bits)



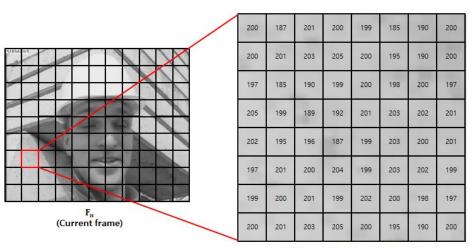








DCT (Discrete Cosine Transform)



8x8 image block(spatial domain)

DC component

DCT

987	223	112	12	1	-1	0	C
201	180	101	23	-4	-1	0	C
14	21	9	5	3	-1	0	C
3	-5	-3	3	-1	-1	0	C
-1	2	4	-1	-1	1	0	C
1	1	-1	1	-1	0	0	C
0	0	1	1	1	0	0	C
0	0	0	0	0	0	0	C

AC components

Transformed block(frequency domain)

source: GitHub - JawThrow/ICSPCodec: Implementation of video codec based on H.261

Two-dimensional DCT & IDCT

$$F(u,v) = \frac{2}{N}C(u)C(v)\sum_{x=0}^{N-1}\sum_{y=0}^{N-1}f(x,y)\cos\frac{(2x+1)u\pi}{2N}\cos\frac{(2y+1)v\pi}{2N}$$

$$f(x,y) = \frac{2}{N}\sum_{u=0}^{N-1}\sum_{v=0}^{N-1}C(u)C(v)F(u,v)\cos\frac{(2x+1)u\pi}{2N}\cos\frac{(2y+1)v\pi}{2N}$$
Where $C(u), C(v) = \begin{cases} \frac{1}{\sqrt{2}} & \text{for } u,v=0\\ 1 & \text{otherwise} \end{cases}$

$$0 \le x, y, u, v \le N-1, N^2 : \text{frame size}$$



Secret: "Hello"



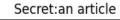
Secret:an article



Secret:image



Secret:image



Decoded_output > ≡ dct_hidden_text.txt

- In the heart of a bustling cit
 - charming bookstore that stood
- It was a haven for book lover:
- with its wooden shelves filled
- The aroma of freshly brewed co
- creating an inviting atmospher
- waiting to be discovered by co
- The owner, an elderly man with
- It was a place where time seen
- in the magic of reading.



Decoded_output > ≡ dct_hidden_text.txt Hello!

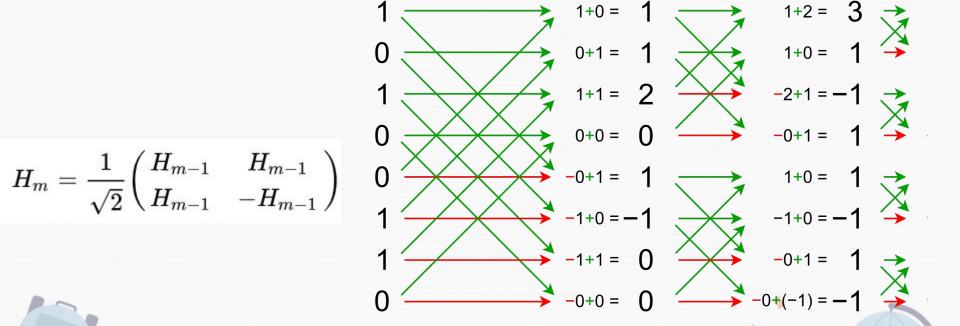


DCT(Discrete Cosine Transform)

Summary of failure reasons:

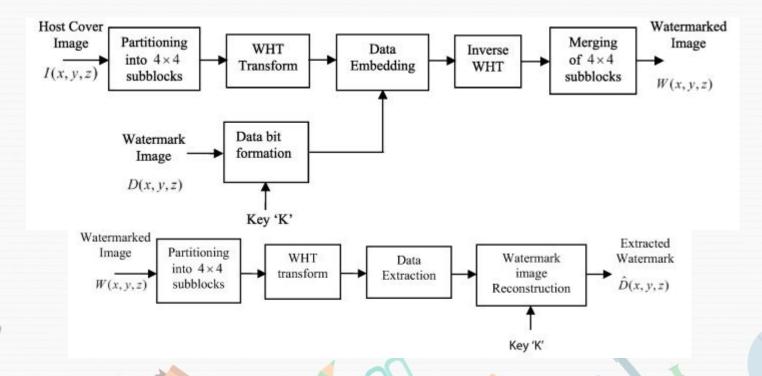
- 1. Floating-point error
 - The volume of text is too large.
- 2. Boundary effects
 - low robustness

WHT

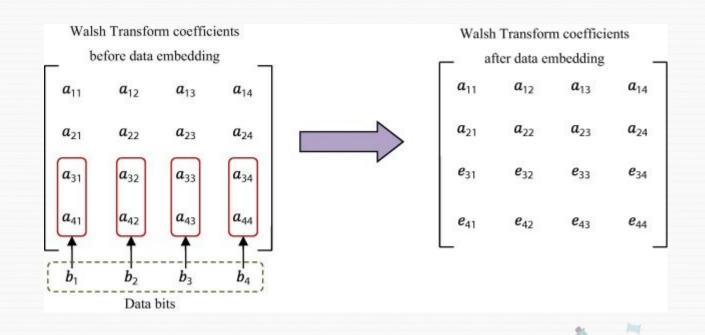


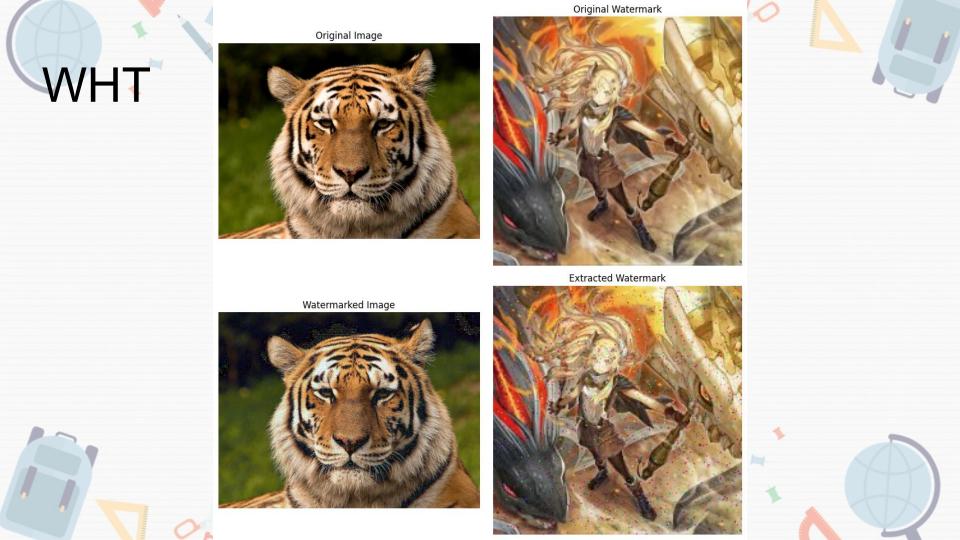
source: https://en.wikipedia.org/wiki/Hadamard_transform

WHT



WHT

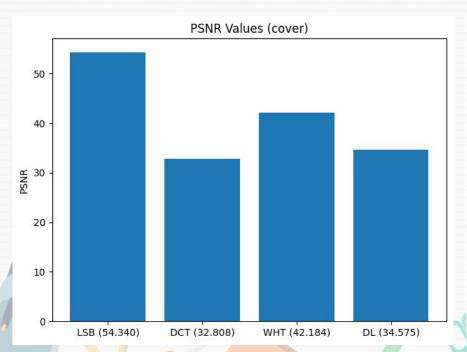




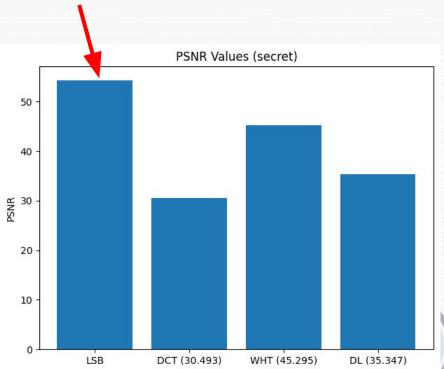
Deep Learning

Evaluation

PSNR

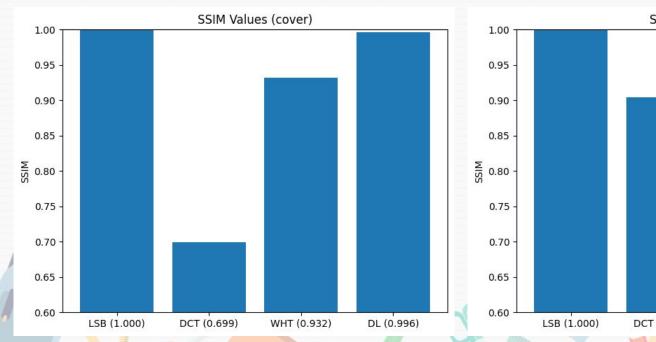


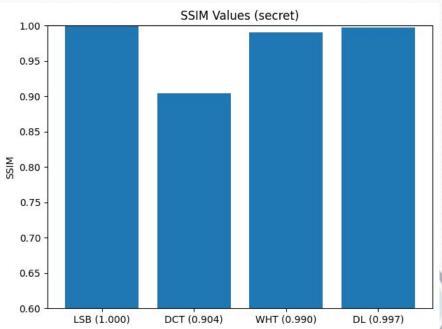
為無限大



Evaluation

SSIM





Comparison

Pros:

- 1. LSB方法能無損精確藏入資訊
- 2. DCT、WHT方法具高隱密性且僅需較低計算量
- 3. 套用神經網路可以實現各種圖片形式間的圖片隱寫術

Cons:

- 1. 若文本量大, 使用DCT方法在Cover藏入資訊後會很明顯
- 2. WHT方法無法避免誤差
- 3. 神經網路架構通常會把input、output設為統一大小

Contribution

- 實作一個成功的LSB方法 (可以藏彩色圖片)
- 實作利用DCT並能藏入圖片的方法(但有諸多限制)
- 透過論文記述公式, 重現並實作WHT方法
- 實驗DL方法可以適用多種影像格式
- 對這四種方法進行數值分析並探討差別

