

Efficient Storage of Images onto DNA Using Vector Quantization

Melpomeni Dimopoulou, Marc Antonini

▶ To cite this version:

Melpomeni Dimopoulou, Marc Antonini. Efficient Storage of Images onto DNA Using Vector Quantization. Data Compression Conference (DCC) 2020, Mar 2020, Utah, United States. 10.1109/DCC47342.2020.00085. hal-02549709

HAL Id: hal-02549709 https://hal.science/hal-02549709v1

Submitted on 21 Apr 2020

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.







Efficient Storage of Images onto DNA Using Vector Quantization

Melpomeni Dimopoulou

Marc Antonini

dimopoulou@i3s.unice.fr

am@i3s.unice.fr

Labratoire 13S, Université Côte d'Azur, CNRS

Biological

Procedures

(channel)

Training

 u_K^{\wedge}

Data Compression Conterence



THE PROBLEM:

- Conventional storage devices can endure for 5-10 years
- 80% of data is cold (very rarely accessed)
- High cost of reliable storage

THE SOLUTION: Use of DNA as a means of digital data storage



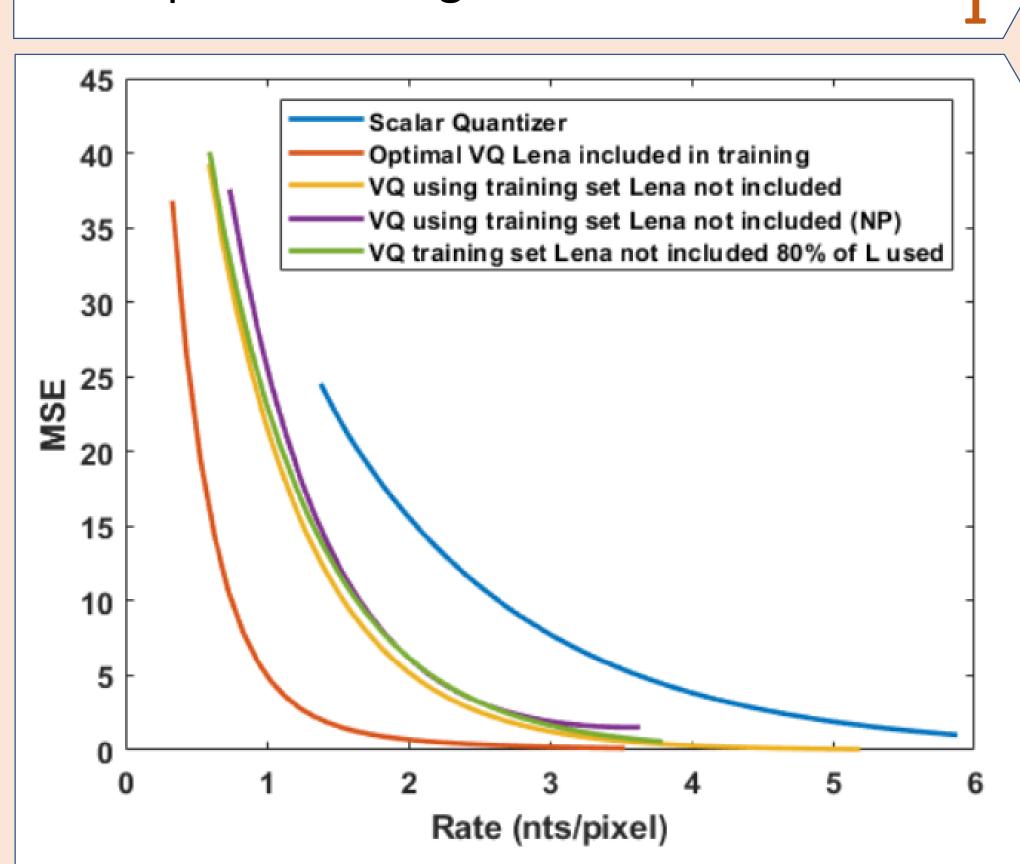
- Longevity
- High capacity
- → Need for efficient encoding to control the cost of DNA synthesis!

PURPOSE OF THIS WORK:

 To propose a new efficient method for the encoding of images into DNA using Vector quantization (VQ) improving the results obtained in our previous work** while controlling the DNA synthesis cost.

THE PROPOSED METHOD:

- DWT decomposition to reduce spatial redundancy
- VQ to encode each DWT subband
- Closed loop source allocation to optimally compress an image



EXPERIMENTAL RESULTS:

- Image: Lena 512x512
- Training set of different face images to obtain the vectors
- NP: No Patterns $\rightarrow L = 2K$
- 80% of L used: K=80% L
- Results using VQ show great improvement compared to the results obtained in our previous work
- ** Dimopoulou, Melpomeni, et al. "A biologically constrained encoding solution for long-term storage of images onto synthetic DNA." 2019 27th European Signal Processing Conference (EUSIPCO). IEEE, 2019.

NUCLEOTIDE ALLOCATION:

Goal:

For a desired rate R_t , find optimal values of k (number of vectors) and n (length of vectors) for VQ that minimizes the distortion.

For each DWT subband:

ENCODER

Encoding

Mapping <

Example of an oligo where

k=L (top oligo) and an oligo

were L=2K (bottom oligo).

The two ways for treating

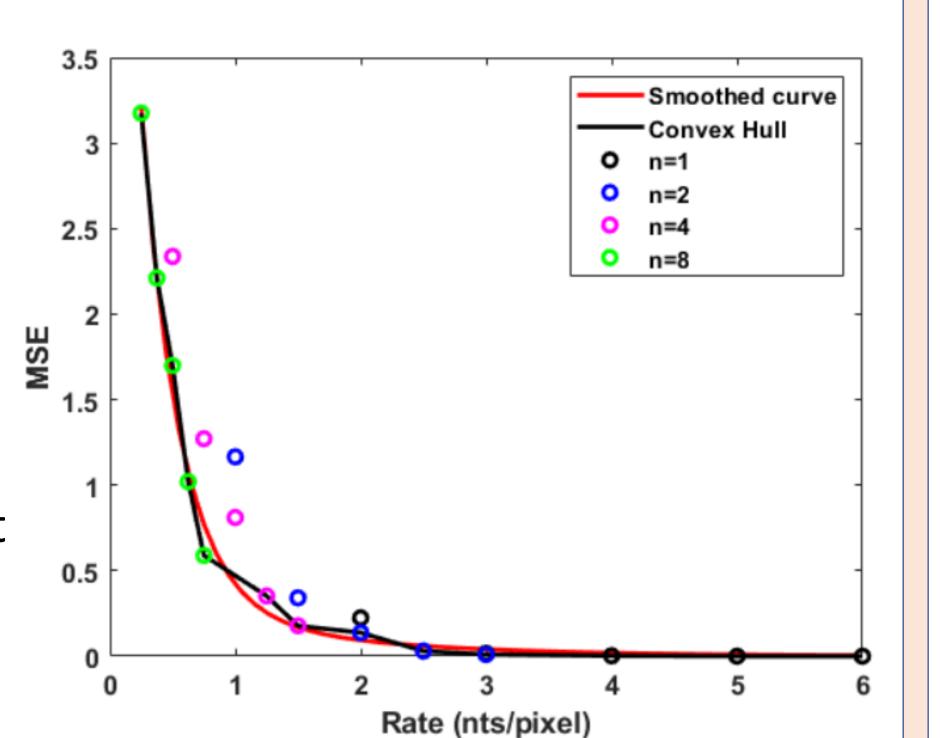
(2) on the right

patterns. (1) on the left and

 $\{A, T, C, G\}$

Formatting -

- Build Rate-Distortion curve for different k and n
- Optimal points are lying on the convex hul



{A,T, C,G}

Deformatting

 $\{A, T, C, G\}$

DNA

The proposed encoding schema

 ν_K^{\wedge}

 c_{p+K}

BIOLOGICAL CONSTRAINTS ON THE ENCODING:

(Reduction of sequencing error)

- 1) No homopolymers
- 2) %G,C ≤ %A,T
- 3) No repetition of short patterns

BUILDING A RESTRICTED QUATERNARY CODE:

Two dictionaries of pair elements:

- $C_1 = \{AT, AC, AG, TA, TC, TG, CA, CT, GA, GT\}$
- $\rightarrow n \ picks$: $L = 10^n \ codewords \ of length l=n*2$
- $C_2 = \{A, T, C, G\}$
- \rightarrow adding a symbol from C_2 at the end of codewords: $L = 10^n * 4$
 - \succ Codewords of an even length l are built by picking n = l/2 pair-elements from C_1
 - \triangleright Codewords of odd length l codewords are built by picking n = (l-1)/2 pair-elements from C_1 and adding a pair element from C_2 at the end

PATTERN REPETITIONS:

- VQ: efficient for compression
- More subband coefficients will be represented by the same vector
- Neighboring coefficients will be encoded to the same codeword \rightarrow pattern repetitions!

○ Solution:

Increase code size L to allow double representation



MAPPING:

DECODER

Lookup

- To avoid patterns we need: $L \geq K$
- Mapping $\Gamma: \Sigma \to C^*$
- m = |L/K|
- $\Gamma(\hat{x}^i) = C^*(i + rand(0, m-1) * K)$

Two ways to treat pattern repetition:

- 1) If $K < L \le 2K$: double representation of most frequent symbols, m=1 (left image)
- 2) If $L \ge 2K$: double representation of every word, m=2 (right image)







