

Image Coding With Neural Network-based Colorization

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Introduction

- Idea: coding solution employing an automatic colorization to recover the chrominance components without transmitting any or just very few colour information
- Motivation:
 - · High performance in the area of automatic colorization of grayscale images
 - · Deep neural networks excel at colorization since it requires semantic understanding of the image





- · Colorization process can be carried out at the decoder after luminance decoding:
 - Deep learning-based colorization network exploits the decoded luminance information to perform
 efficient luma-to-chroma estimation
 - Extraction, coding and transmission of some color information is needed to obtain accurate colorizations

Objective and High-level Framework

Proposes and evaluates the first image coding solution adopting a deep learning-based colorization process to recover the chrominance components at the decoder

 To remove color ambiguity and boost color fidelity, chrominance values for specific pixel locations (color hints) are sent by the encoder to the decoder



 Methods are proposed for the selection of the color hints at the encoder, targeting the increase of the colorization performance

Real-Time User-Guided Colorization

- The state-of-the-art solution on user-guided colorization has been adopted since it is the best available:
 - R. Zhang, J. Zhu, P. Isola, X. Geng, A. Lin, T. Yu, A. Efros, "Real-Time User-Guided Image Colorization with Learned Deep Priors," in ACM Transactions on Graphics, vol. 36(4), 119:1-119:11, 2017.
- · Automatic colorization process with the capability of including color information from user inputs
- Can be adapted to include original image color information for some locations



Deep Learning Colorization-based Image Coding (DLCIC)

- Allows to reduce the bit rate associated to the coding of color information
- Includes an intelligent automatic selection of color hints, i.e. ground truth color information



Image Segmentation

 Simple Linear Iterative Clustering (SLIC) algorithm to cluster the image pixels based on their spatial proximity and color similarity



Color Hint Selection

- Several intelligent and efficient color hint selection methods are proposed, from a ratedistortion performance point of view
- No Hints, where no hints are extracted and the colorization obtained at the decoder is only based on the decoded luminance
- Random hints, location of the color hints is chosen randomly, thus ignoring the image segmentation sub-module, following a Normal distribution located at the image center
- Region-driven hints, makes use of the regions created by the Image Segmentation module to extract more appropriate color hints depending on the image content.
- For each segmented region, a hint location is selected corresponding to the pixel with the least color difference to all the remaining pixels in the region.
- Colorization error-driven hints, hints are selected according to the colorization error, i.e. hints are located in the areas where the colorization error is higher

· Can be totally independent from the previous image segmentation or use it for bootstrapping



Test Conditions

- The proposed DLCIC solutions (using different hint extraction methods) are evaluated in terms of RD performance and subjective quality:
 - · Four test images from the JPEG AI dataset with different characteristics
- Peak Signal-to-Noise Ratio (PSNR) YCbCr, Structural Dissimilarity Index Measure (DSSIM)
 Coding solutions evaluated:
- JPEG with a 4:2:0 color subsampling format (JPEG 420)
- · JPEG 400 NH (no hints): colorization process only based on the decoded luminance
- JPEG 400 RH (random hints): colorization process based on the decoded luminance and some spatially random generated hints
- JPEG 400 R-dH LD (region-driven hints), color hints locations correspond to pixels from each segmented color-uniform region of the image
- JPEG 400 CE-dH SC-5px (colorizer error-driven hints), color hints locations are selected based on the colorization error with a minimum distance between hints of 5 pixels to avoid collapsing many hints in a small area
- For all the proposed DLCIC solutions, only the luminance is JPEG coded, varying the quality factors between 5% to 90% and 1000 color hints are considered.

Experimental Results

Table 1: BD-PSNR for the proposed DLCIC coding solutions (using as reference the IPEG 420 heni-hmark).

| DLCIC SOLUTION | BD-PSNR YCbCr [dB] | | | |
|-----------------------|--------------------|----------------|-----------|--------|
| (1000 HINTS) | Woman | Footballplayer | Buddhists | Harley |
| JPEG 400 NH | -3.72 | -2.35 | -1.99 | -1.94 |
| JPEG 400 RH | -0.96 | -1.78 | -1.61 | -0.55 |
| JPEG 400 R-dH LD | -0.68 | -1.17 | -1.26 | -0.46 |
| JPEG 400 CE-dH SC-5px | -0.57 | -0.89 | -1.12 | -0.34 |



Figure 3: Comparative results for subjective inspection for Woman (top row) and Footballplayer (bottom row) test images

Conclusions

Objective quality assessment:

- The JPEG 400 CE-dH SC-5px solution always outperforms the other DLCIC solutions
- The colorization performance improvement with 'increased intelligence' color hints extraction methods is larger for Footballplayer image and lower for the Harley image
- All DLCIC solutions perform worse than JPEG 420
- Subjective quality assessment:
- · Promising subjective quality is obtained with the DLCIC coding solutions
- · The lowest (decoded) quality is achieved with the JPEG 400 NH solution
- The best (decoded) quality is achieved using 'more intelligent' and content aware DLCIC coding solutions in terms of colorization, as it can be observed for Footballplayer
- Objective quality metrics do not correlate well with perceived quality, especially since colorizations are not pixel-level faithful even if they are subjectively acceptable
- · Future work must include formal subjective quality assessment