

INTRODUCTION

- For security reasons, more and more images are transferred or stored in encrypted domains by using selective format-compliant JPEG encryption methods.
- It is necessary to evaluate the confidentiality of the selective crypto-compressed JPEG images.
- Quality metrics, such as PSNR or SSIM, give a very low correlation with a mean opinion score (MOS) for low quality images.
- We propose an efficient confidentiality metric based on the visual saliency diffusion.
- We show experimentally that this metric is well correlated with a MOS and efficient to evaluate the confidentiality of selective crypto-compressed JPEG images.

SELECTIVE CRYPTO-COMPRESSION

- Crypto-compression targets JPEG images.
- We have several parameters to select:
 - Crypto-compression algorithm : FIBS or SJCC
 - Encryption of DC and/or AC
 - Encryption of Luminance and/or Chrominance
- 27 relevant combinations
- FIBS**: full inter-block shuffle [1]. This method scrambles DC coefficients as well as same frequency AC coefficients.
- SJCC**: a selective JPEG crypto-compression method [2]. This method encrypts the amplitude part of non null AC coefficients of each block and changes the DCT coefficients histogram.

200 images from the BSD500 dataset as input images for a total of 5400 crypto-compressed images www.lirmm.fr/~wpuech/dataset

EVALUATION OF THE CRYPTO-COMPRESSED JPEG IMAGES

- MOS: arithmetic mean of ratings given by humans for a particular stimulus.
 - a single number, from 1 to 5, used to describe the quality of the current stimulus, where 5 is the best score and 1 is the worst.
- Evaluation on 41 different people, male and female from 17 to 53 years old:
 - 1: The distortion is unbearable, nothing is visible**
 - 2: The distortion is very annoying, I can barely make-out the content**
 - 3: The distortion is annoying, but I can see the content**
 - 4: The distortion is slightly annoying, but the content is clear**
 - 5: The distortion is not annoying at all**

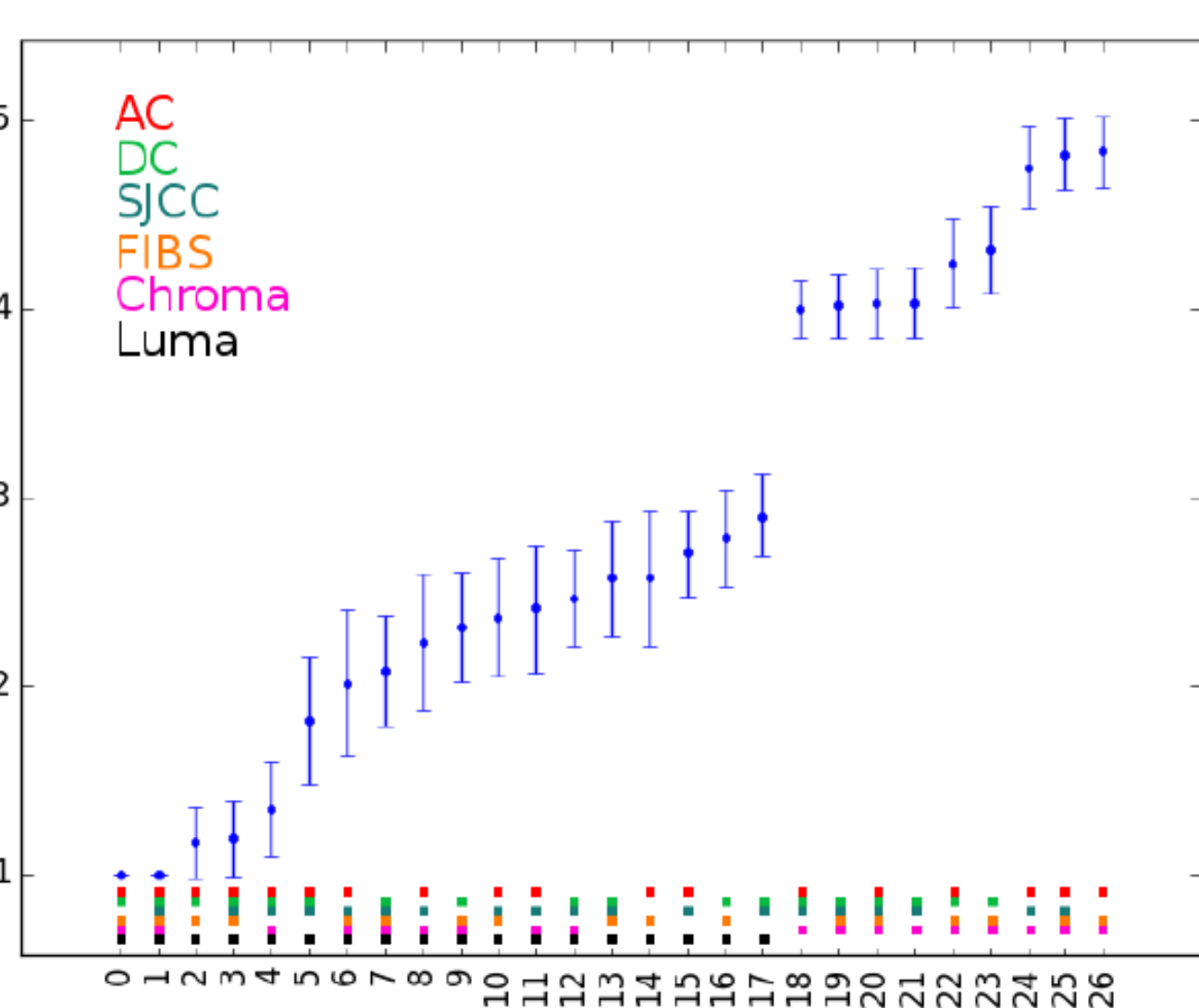


Fig. 2: MOS for the 27 distortions.



Fig. 1: An example of images using different selective crypto-compression methods with their corresponding MOS.

- PSNR: PSNR does not interact well with human judgment, The range is $[0; +\infty]$, where identical images have a PSNR of $+\infty$.
- SSIM: (Structural Similarity Measure). A luminance, a contrast and a structure score are combined. The range is $[0;1]$ where identical images have a score of 1.
- ESS [3]: (Edge Similarity Score). It uses non overlapping 8×8 block directions. With the range $[0;1]$, a higher score reflects a less distorted image.
- LSS [3]: (Luminance Similarity Score). It uses non overlapping 8×8 block average luminance. With the range $[-8.5; 1]$ for default parameters of $\sigma = 0.1$ and $\tau = 3$, a higher score reflects a less distorted image.
- NPCR: the number of pixel changes between images. Its range is $[0;100]$, where a fully encrypted image has a NPCR close to 100, where almost all the pixels have been changed.
- UACI: the unified averaged changed intensity. It is the average intensity difference between two images. Its range is $[0;100]$, where a fully encrypted image has a value close to 33.

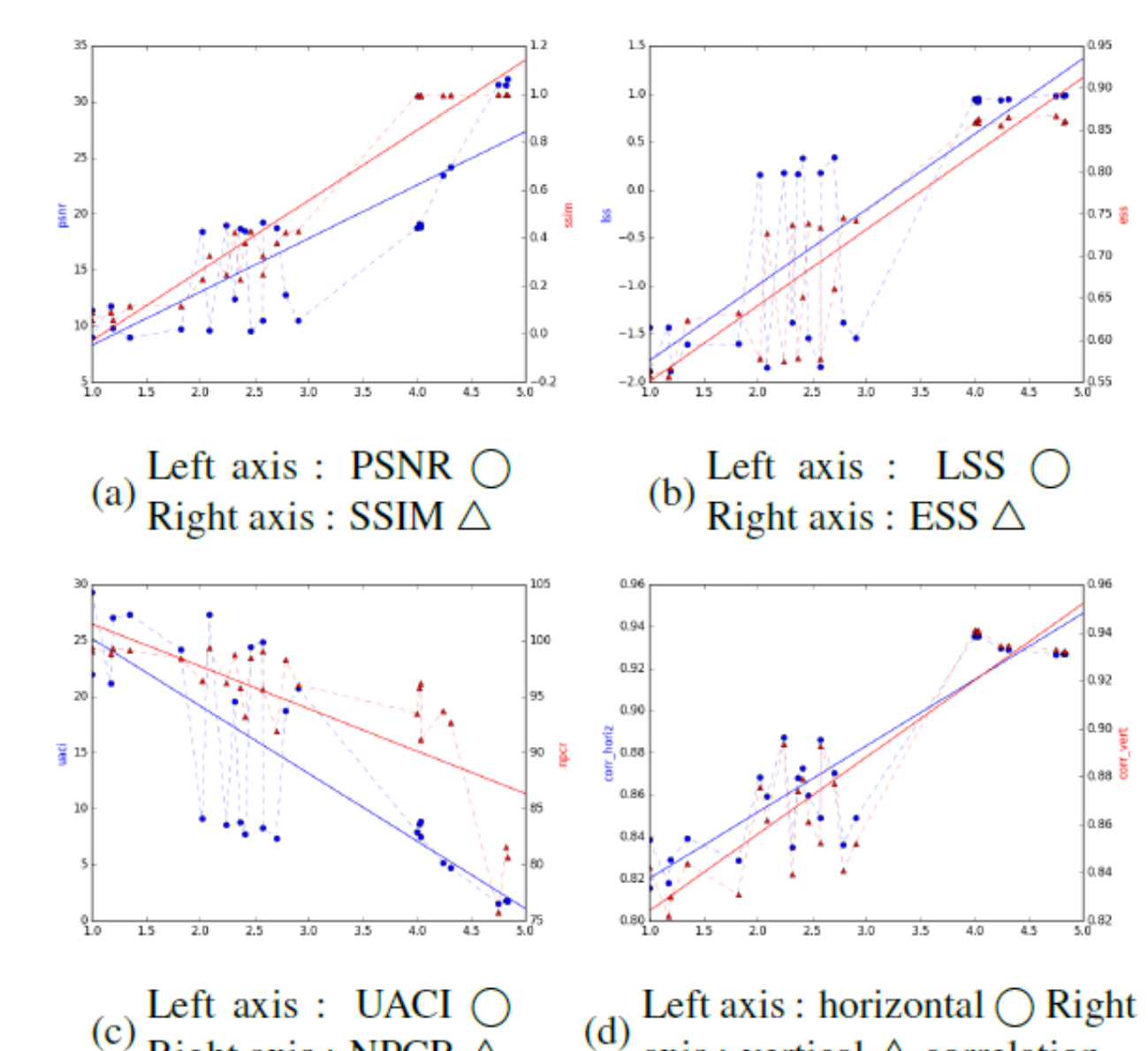


Fig. 3: Plots of different metrics with the MOS on the x-axis, a: PSNR and SSIM, b: LSS and ESS, c: UACI and NPCR, d: horizontal and vertical correlation.

VISUAL SALIENCY-BASED CONFIDENTIALITY METRIC

- Our proposed metric:
- Important information is located in salient areas: if no salient areas can be found, then the content is hidden.
- Let M_o be the saliency map of the original image and M_c be the saliency map of the crypto-compressed image.
- Two binary images are thus created, B_o from M_o (Fig. 4b) and B_c from M_c (Fig. 4e). The score based on the visual saliency is then:
$$v_{saliency} = \frac{\sum_{i=0}^{width} \sum_{j=0}^{height} B_o(i,j) \times B_c(i,j)}{\sum_{i=0}^{width} \sum_{j=0}^{height} B_o(i,j)}$$
- The results however are not as good for mid quality images, when the MOS is around 2 and 3.
- We introduce a second score, v_{edges} , based on the Sobel operator in an attempt to stabilize our first score $v_{saliency}$ (Fig. 4c and Fig. 4f).
- The final score is (with $\alpha = 0.6$):
$$v = \alpha * v_{saliency} + (1 - \alpha) * v_{edges}$$
- The euclidean distance of our metric to the MOS of all the distortions is 0:4323 for our metric and 0:5095 for SSIM and the euclidean distance of our metric to individual image rating is 0:6674 and 0:6699 for SSIM

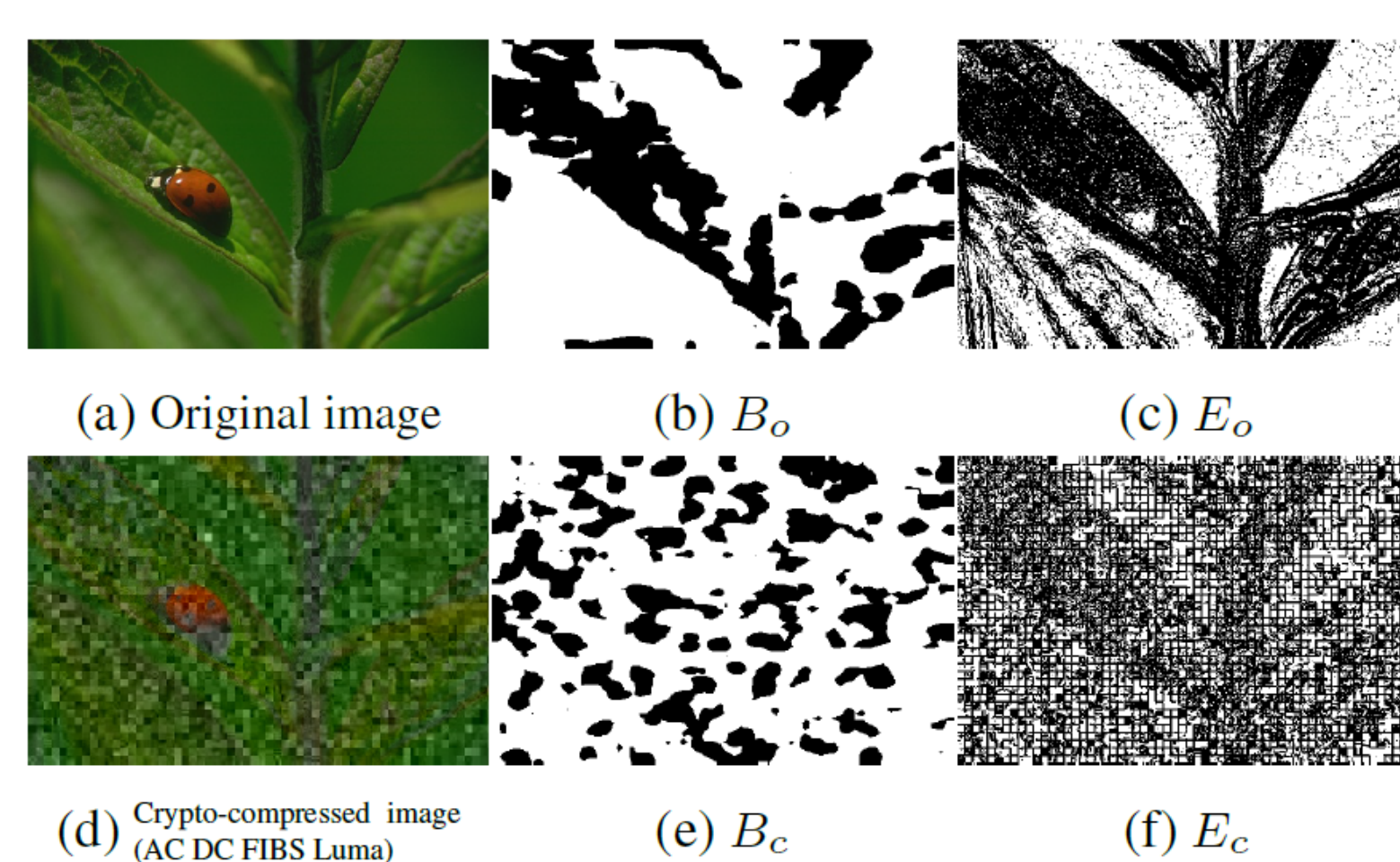


Fig. 4: Maps obtained for an original image a), a crypto-compressed image d), b) and e) saliency maps, c) and f) edge maps.

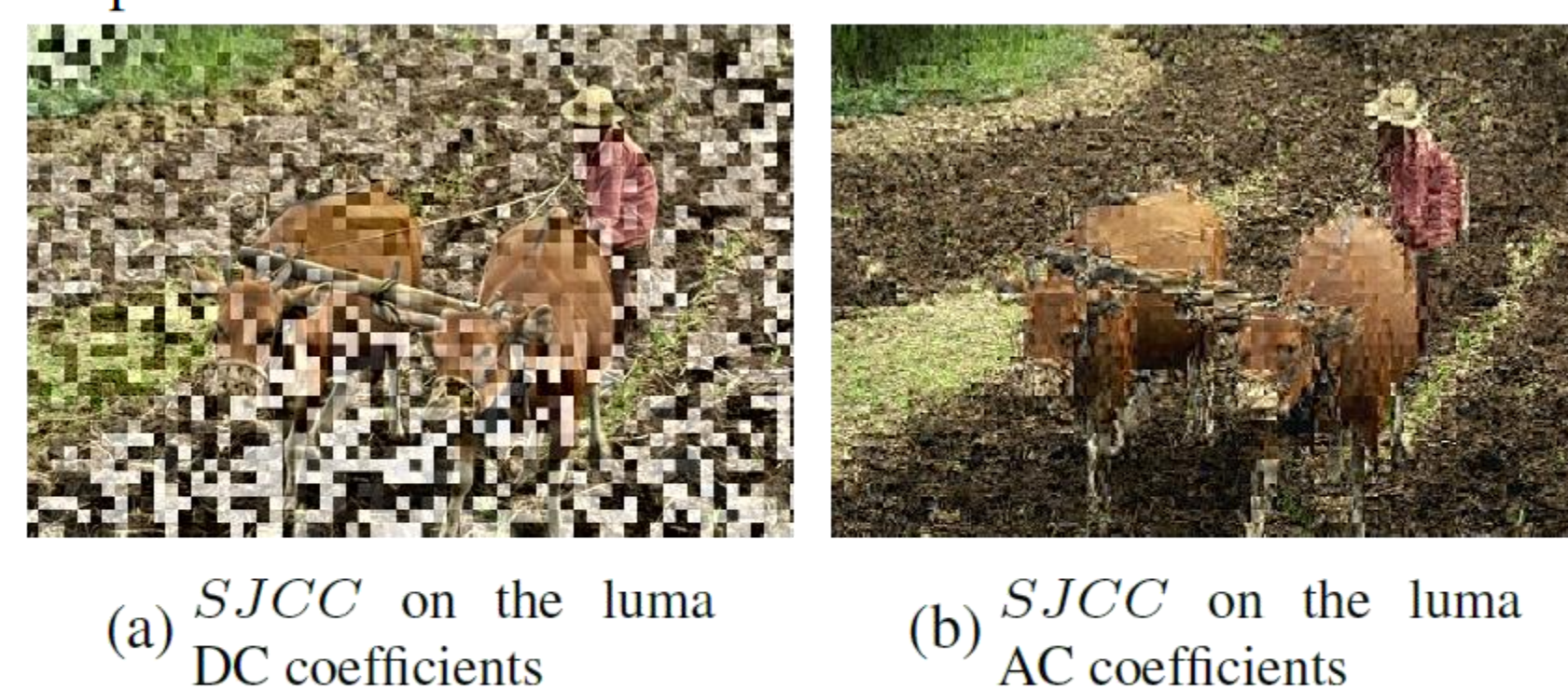


Fig. 5: Global noise caused by a SJCC on DCT coefficients.

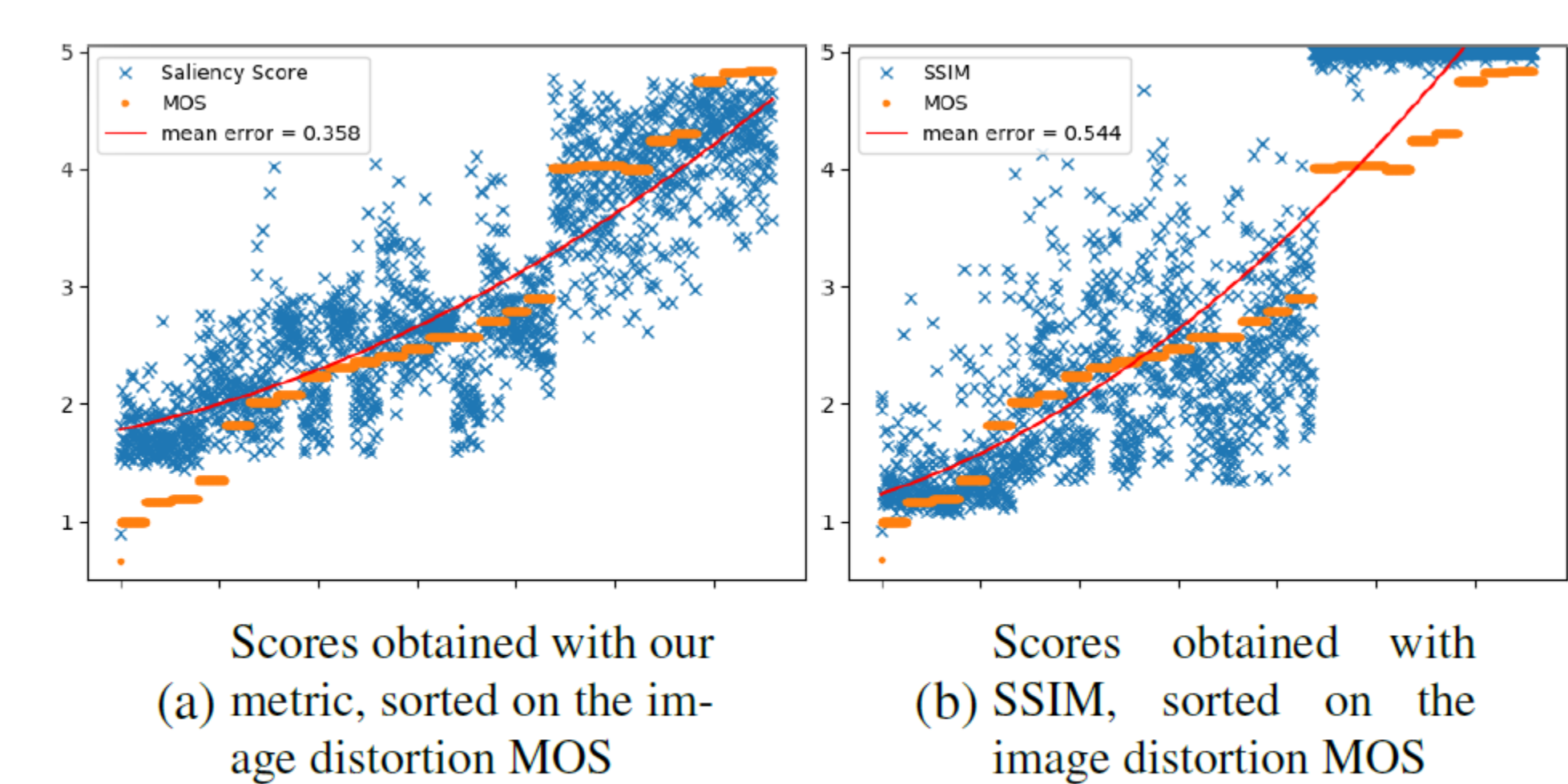


Fig. 6: Comparison of results for our metric and SSIM.

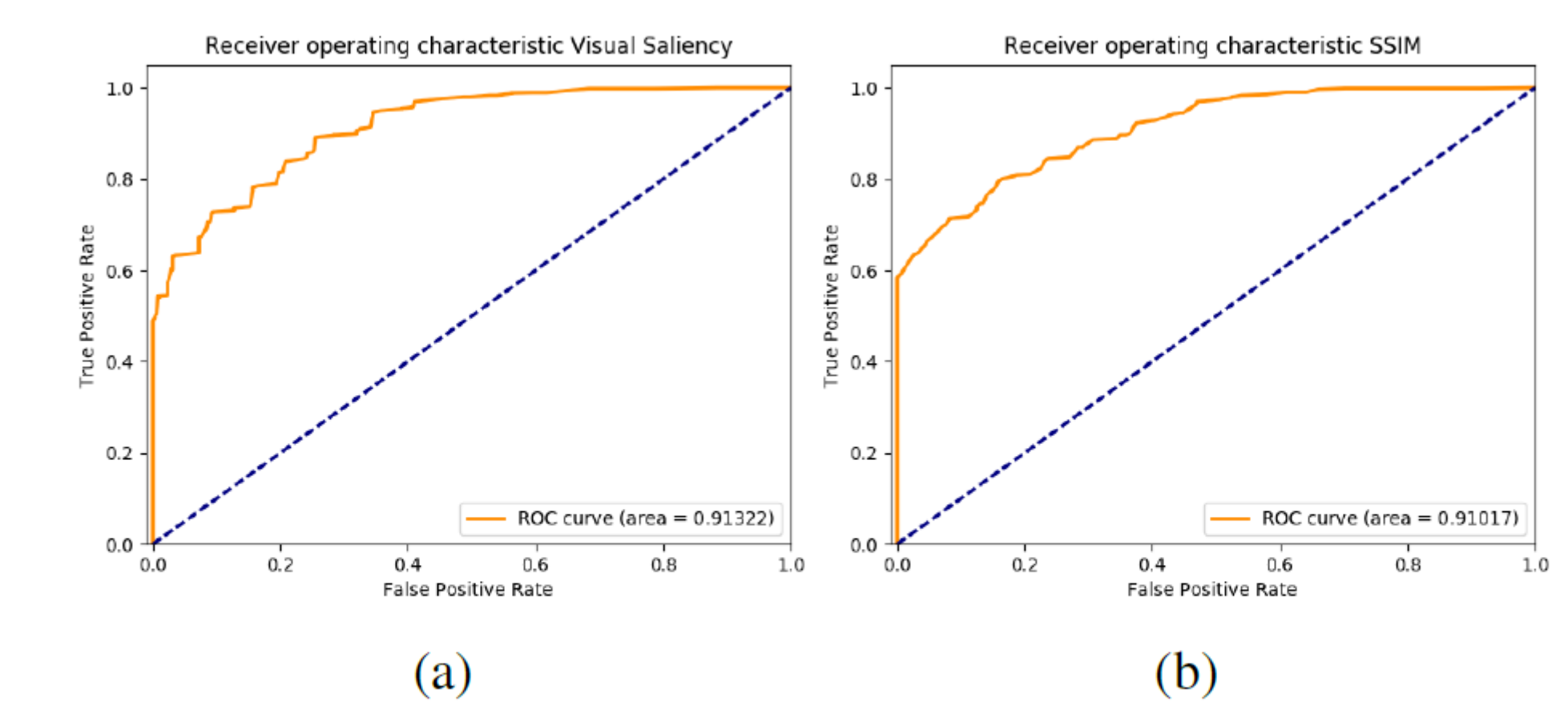


Fig. 7: ROC curves of: a) Our metric, b) SSIM.

CONCLUSION

- We proposed a dataset composed of selective crypto-compressed images for image quality assessment.
- The images were rated by human observers to obtain a mean opinion score.
- We introduced a new confidentiality metric based on visual saliency.
- We evaluated our metric and noticed that we obtained better results compared to quality metrics such as SSIM.
- Future work:
 - A more in depth analysis of our dataset and each of its 27 distortions,
 - A more refined metric based on visual saliency, which shows great potential.

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