

# Users prefer Jpegli over same-sized libjpeg-turbo or MozJPEG

Martin Bruse\*, Luca Versari†, Zoltan Szabadka†, Jyrki Alakuijala†

## Abstract

We performed pairwise comparisons by human raters of JPEG images from MozJPEG[1], libjpeg-turbo[2] and our new Jpegli[3] encoder. When compressing images at a quality similar to libjpeg-turbo quality 95, the Jpegli images were 54% likely to be preferred over both libjpeg-turbo and MozJPEG images, but used only 2.8 bits per pixel compared to libjpeg-turbo and MozJPEG that used 3.8 and 3.5 bits per pixel respectively.

The raw ratings[4] and source images are publicly available for further analysis and study.

## Introduction

We created a new JPEG encoder and decoder, Jpegli, both API-, and ABI-compatible with libjpeg, MozJPEG, and libjpeg-turbo. The new encoder uses an adaptive dead-zone quantization technique, and quantization matrices tuned for combined Butteraugli[5] and SSIMULACRA[6] metrics. We studied which encoder output human evaluators prefer.

The settings were based on the methodology used by CLIC 2024[7]: Pairwise comparisons between images, both in standard JPEG format, to see which one human raters prefer. A publicly available[8] procedure was used to compute the Elo[9] score per method that best explains the results.

## Materials and Methods

### Source images

To ensure a meaningful evaluation of compressor quality and output size, we selected the validation set of the publicly available CID22[10] dataset from Clouldinary, consisting of 49 images with known source and processing. The images were chosen to cover a wide range of contents, and are composed of a mixture of people, objects, scenery, and graphical elements.

### Image degradation method

We tested three encoders which offer a distortion vs size trade-off via a single quality parameter and a chroma subsampling setting. We varied the quality parameter between 55 and 95 for Jpegli and libjpeg-turbo, and between 70 and 95 for MozJPEG, and vary the chroma subsampling setting between YUV420 and YUV444 for Jpegli, and between YUV420, YUV422, and YUV444 for MozJPEG and libjpeg-turbo.

Each combination of encoder, quality parameter, and chroma subsampling setting defines a degradation method.

The Jpegli degraded images were produced using the Linux shell command `cjpegli input.png output.jpeg --quality [QUALITY] --chroma_subsampling=[444|420]`, and the MozJPEG- and libjpeg-turbo-degraded images were produced using the command `env LD_LIBRARY_PATH=$HOME/[mozjpeg|libjpeg-turbo]/build/ convert input.png -quality [QUALITY] -sampling-factor [4:4:4|4:2:2|4:2:0] output.jpeg`.

---

\*Google Research, Kungsbron 2, 111 22 Stockholm, Sweden

†Google Research, Brandschenkestrasse 110, 8002 Zürich, Switzerland

## Viewing environment

The human evaluators viewed the images under their own normal viewing preferences - the experiment controlled neither lighting conditions nor computer screens or display settings. As such, the ratings produced by this experiment are expected to align well with in-the-field use of the JPEG libraries under study.

## Experiment design

We chose a pairwise comparison model, and modeled the probability that a human evaluator would prefer each method over each other method as a random variable presented as an Elo score.

We chose pairs of images to compare based on which result would reduce the uncertainty of the Elo score the most. The human evaluators were shown the original image along with one of the degraded images, and were allowed to replace the degraded image with the other as many times as necessary to make the decision of which degraded image is closest to the original. The evaluators were forced to make a decision for each presented pair. This approach leads to methods with difficult to distinguish images having Elo scores close together.

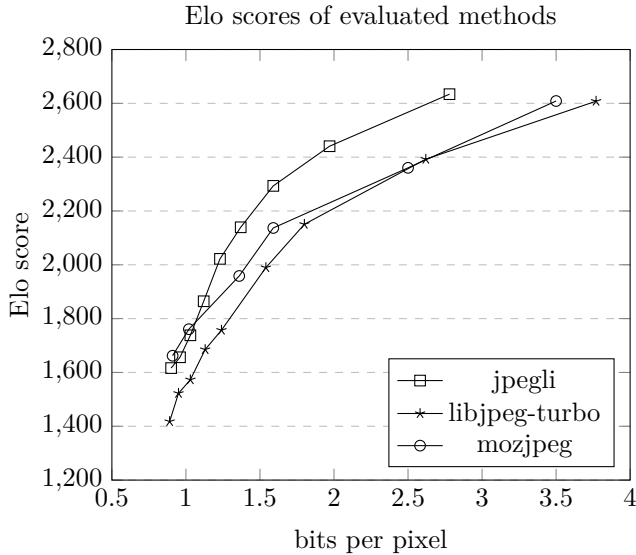
To ensure evaluator quality, we seeded the images with ‘golden’ questions where one of the degraded images was the original image and the other a heavily degraded image, e.g. a JPEG image with quality 50, and prevented assigning more questions to evaluators that provided the wrong answer to multiple ‘golden’ questions.

## Subjects

18 raters participated in our experiment. Subjects were recruited from an internal Google team of human raters specializing in providing high-quality data to machine learning systems. Subjects were not specially selected for age, gender, experience, or background.

## Results

Each subject generated between 86 and 995 answers during 58 days, with a mean of 749, a median of 912, and a standard deviation of 340. The results were analyzed by computing the Elo scores that best explain the evaluated results, following the CLIC 2024 methodology.



Overall, for bitrates above 1 bit per pixel the evaluators were always more likely to prefer jpegli images to libjpeg-turbo or MozJPEG at similar bitrates. The computed Elo scores, their 99% credible intervals, and mean bits per pixel is listed in Appendix A.

## Discussion

Our headline result is that Jpegli produces JPEGs preferred by human raters at a lower bitrate using a library that is both API- and ABI-compatible with libjpeg-turbo and MozJPEG.

To simplify comparison of bitrates at normalized quality levels, we performed a linear interpolation of the bitrates at each Elo score for the compared methods. The code used is publicly available[11]. The results of this interpolation for the Elo scores of each evaluated libjpeg-turbo setting is listed in Appendix B.

Our results show clear evaluator preference for Jpegli with bitrates over 1 bit per pixel, which according to the HTTP Archive’s Web Almanac 2022[12] accounts for 75% of all images found on the internet. According to the same source, the median bitrate for JPEG images is 2.1 bits per pixel. According to our interpolation libjpeg-turbo at 2.1 bits per pixel corresponds to an Elo of approximately 2238, while Jpegli at Elo 2238 corresponds to a bitrate of approximately 1.5 bits per pixel, or a reduction of about 28%.

## Conclusion

In a pairwise comparison by 18 human raters of Jpegli, libjpeg-turbo, and MozJPEG encodings of 49 publicly available images, computed Elo scores show that at comparable bitrates jpegli encodings are likelier to be preferred at rates over 1 bit per pixel. We conclude that Jpegli generates the same or higher quality images at the same or lower bitrates.

## References

- [1] Mozilla Foundation. Mozilla JPEG Encoder Project 2014. <https://github.com/mozilla/mozjpeg>.
- [2] libjpeg-turbo contributors. libjpeg-turbo 2015. <https://libjpeg-turbo.org/>.
- [3] Jpegli contributors. Jpegli 2022. <https://github.com/libjxl/libjxl/tree/main/lib/jpegli>.
- [4] Versari L, Szabadka Z, Bruse M, Alakuijala J. mucped23 answers 2023. <https://github.com/google-research/google-research/blob/master/mucped23/answers.csv>.
- [5] Butteraugli contributors. Butteraugli 2016. <https://github.com/google/butteraugli>.
- [6] SSIMULACRA contributors. SSIMULACRA 2017. <https://github.com/cloudinary/ssimulacra>.
- [7] Ballé J, Toderici G, Versari L, Johnston N, Theis L, Norkin A, et al. CLIC 2024 2024. <https://compression.cc/>.
- [8] Versari L. ELO-based pairwise comparison system with rater modeling 2024. [https://github.com/google-research/google-research/tree/master/elo/\\_rater/\\_model](https://github.com/google-research/google-research/tree/master/elo/_rater/_model).
- [9] Wikipedia contributors. Elo rating system 2024. [https://en.wikipedia.org/wiki/Elo/\\_rating/\\_system](https://en.wikipedia.org/wiki/Elo/_rating/_system).
- [10] Sneyers J, Ben Baruch E, Vaxman Y. Cloudinary Image Dataset ’22 2022. <https://cloudinary.com/labs/cid22>.
- [11] Bruse M. Quality interpolation Python notebook 2024. [https://github.com/google-research/google-research/blob/master/mucped23/equivalent/\\_quality/\\_interpolation.ipynb](https://github.com/google-research/google-research/blob/master/mucped23/equivalent/_quality/_interpolation.ipynb).
- [12] Web Almanac ’22, Media 2022. <https://almanac.httparchive.org/en/2022/media>.

## Appendix A

method	elo	p99Low	p99Hi	bpp
jpegli-q55-yuv444	1616.22	1570.12	1662.31	0.9
jpegli-q60-yuv444	1656.44	1611.1	1701.78	0.96
jpegli-q65-yuv420	1600.83	1554.37	1647.3	0.87
jpegli-q65-yuv444	1738.07	1695.75	1780.39	1.03
jpegli-q70-yuv420	1692.46	1648.48	1736.43	0.95
jpegli-q70-yuv444	1864.52	1824.31	1904.73	1.12
jpegli-q75-yuv420	1823.93	1782.02	1865.83	1.05
jpegli-q75-yuv444	2022.06	1980.59	2063.53	1.23
jpegli-q80-yuv420	1980.54	1935.38	2025.7	1.18
jpegli-q80-yuv444	2139.41	2096.15	2182.67	1.37
jpegli-q85-yuv420	2135.4	2088.14	2182.65	1.36
jpegli-q85-yuv444	2293.27	2247.15	2339.39	1.59
jpegli-q90-yuv420	2296.94	2247.53	2346.35	1.7
jpegli-q90-yuv444	2440.64	2388.39	2492.89	1.97
jpegli-q95-yuv420	2481.99	2425.26	2538.71	2.41
jpegli-q95-yuv444	2634.02	2568.28	2699.77	2.78
libjpeg-turbo-q55-yuv420	1417.72	1359.51	1475.93	0.89
libjpeg-turbo-q60-yuv420	1522.56	1472.45	1572.67	0.95
libjpeg-turbo-q65-yuv420	1572.97	1525.96	1619.99	1.03
libjpeg-turbo-q70-yuv420	1685.13	1642.59	1727.67	1.13
libjpeg-turbo-q75-yuv420	1757.05	1714.45	1799.65	1.24
libjpeg-turbo-q80-yuv422	1989.65	1947	2032.29	1.54
libjpeg-turbo-q85-yuv422	2150.56	2107.5	2193.62	1.8
libjpeg-turbo-q90-yuv444	2392.53	2341.05	2444.02	2.62
libjpeg-turbo-q95-yuv444	2608.02	2546.63	2669.4	3.77
mozjpeg-q70-yuv420	1662.13	1619.22	1705.03	0.91
mozjpeg-q75-yuv420	1760.71	1719.52	1801.9	1.02
mozjpeg-q80-yuv422	1958.47	1916.88	2000.06	1.36
mozjpeg-q85-yuv422	2136.87	2093.91	2179.82	1.59
mozjpeg-q90-yuv444	2360.31	2312.13	2408.48	2.5
mozjpeg-q95-yuv444	2608.61	2546.69	2670.53	3.5

## Appendix B

libjpeg_turbo_equiv_quality	elo	libjpeg_turbo_bitrate	mozjpeg_bitrate	jpegli_bitrate
libjpeg-turbo-q70-yuv420	1685.13	1.13	0.94	0.99
libjpeg-turbo-q75-yuv420	1757.05	1.24	1.02	1.04
libjpeg-turbo-q80-yuv422	1989.65	1.54	1.40	1.21
libjpeg-turbo-q85-yuv422	2150.56	1.80	1.65	1.39
libjpeg-turbo-q90-yuv444	2392.53	2.62	2.63	1.85
libjpeg-turbo-q95-yuv444	2608.02	3.77	3.50	2.67