Appeal and quality assessment for Al-generated images

Steve Göring, Rakesh Rao Ramachandra Rao, Rasmus Merten, Alexander Raake

Audiovisual Technology Group; Technische Universität Ilmenau, Germany Email: [steve.goering, rakesh-rao.ramachandra-rao, rasmus-leo-lukas.merten, alexander.raake]@tu-ilmenau.de

Code & Data: http://git.avt-imt.de/avt_ai_images

Introduction

- ▶ increase of Al-generated images, e.g.:
 - DALL-E-2, Midjourney, Stable Diffusion [7], or Craiyon [1]
- \blacktriangleright text prompt \rightarrow generated image (=text to image (T2I))
- ▶ example images, see Fig. 1
 - text prompt "Hyper-realistic photo of an abandoned industrial site during a storm" (p16)
- uncommon artificial-looking distortions, varying appeal visual quality
- published AVT-AI-Image-Dataset [3]: • appeal, realism, text prompt matching \circ 5 T2l generators





- related work: usually no comparison of several generators
- open: image quality and appeal

Overview of the AVT-AI-Image-Dataset

- ► AVT-AI-Image-Dataset: 27 text prompts, 16 from Drawbench [8]
- ▶ 11 real images included (p17 to p27); all images: resolution 512x512
- ▶ 146 images, full overview in [3], prompt selection see:

ID Prompt	Origin
p11 A mechanical or electrical device for measuring time	Drawbench
p16 Hyper-realistic photo of an abandoned industrial site during a storm	Drawbench
p20 Purple flowers with yellow and a small bug	own
p23 A portrait of a mule	own
p27 A box with tools for home office	own

Subjective Test Design and Evaluation

- ▶ similar to [4, 6, 3]; AVRate Voyager [2] with two 1-5 sliders

Figure 1: Generated images for p16: DALL-E-2 (left), Midjourney (right).

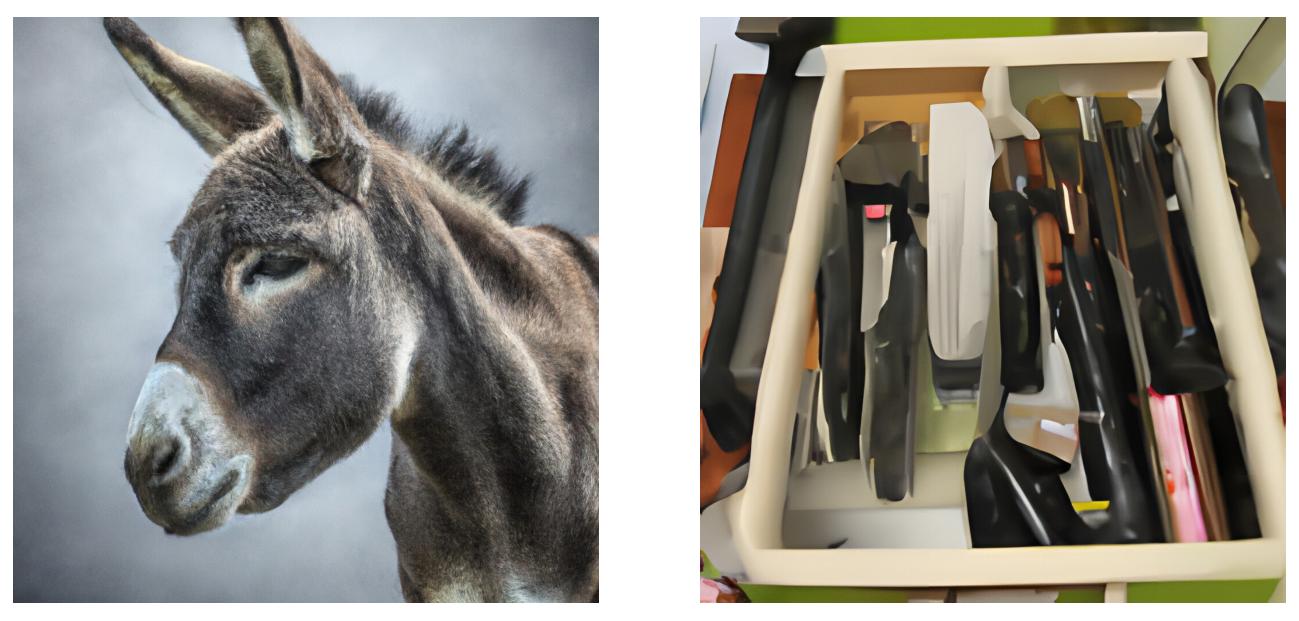


Figure 2: Best quality (left): DALL-E-2, p23 and worst (right): Glide, p27.

Evaluation of Image Quality

- ▶ SOS-analysis [5]; *a* value ≈ 0.306
- ▶ Midjourney, DALL-E-2 best, see Fig. 4
- ▶ 25 participants (12 from clickworker.com, remaining from university)
- \blacktriangleright no training phase, \approx 30 mins; partial runs excluded in results

Evaluation of Image Appeal

- ► SOS-analysis [5]; *a* value ≈ 0.33
- \blacktriangleright cross-test comparison: Pearson \approx 0.91, Kendall \approx 0.75, Spearman \approx 0.9
- ▶ highest: Midjourney p16; lowest: Glide p11; compare Fig. 3

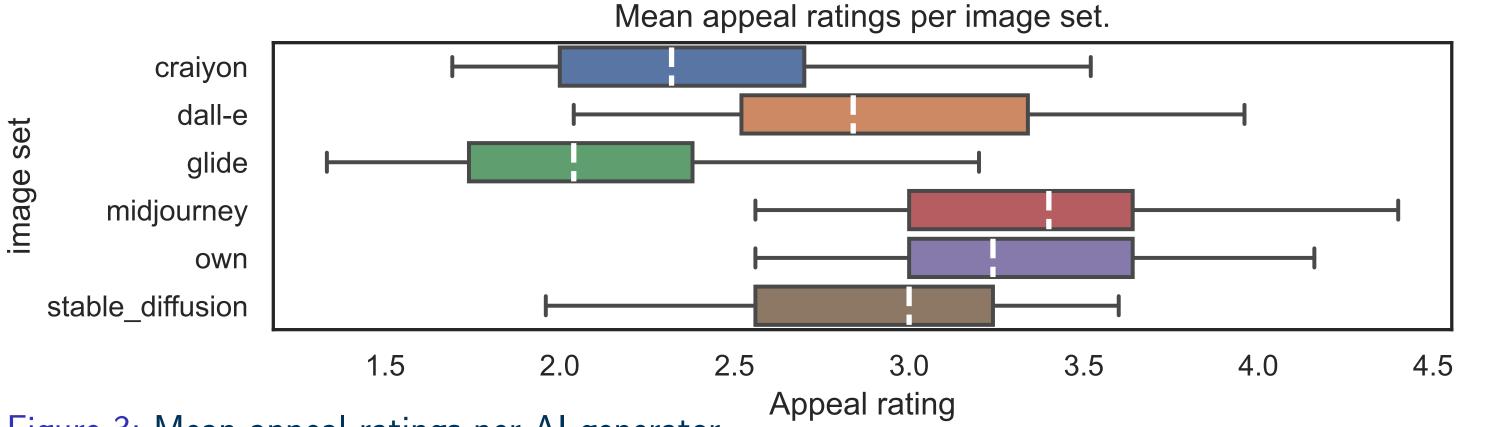
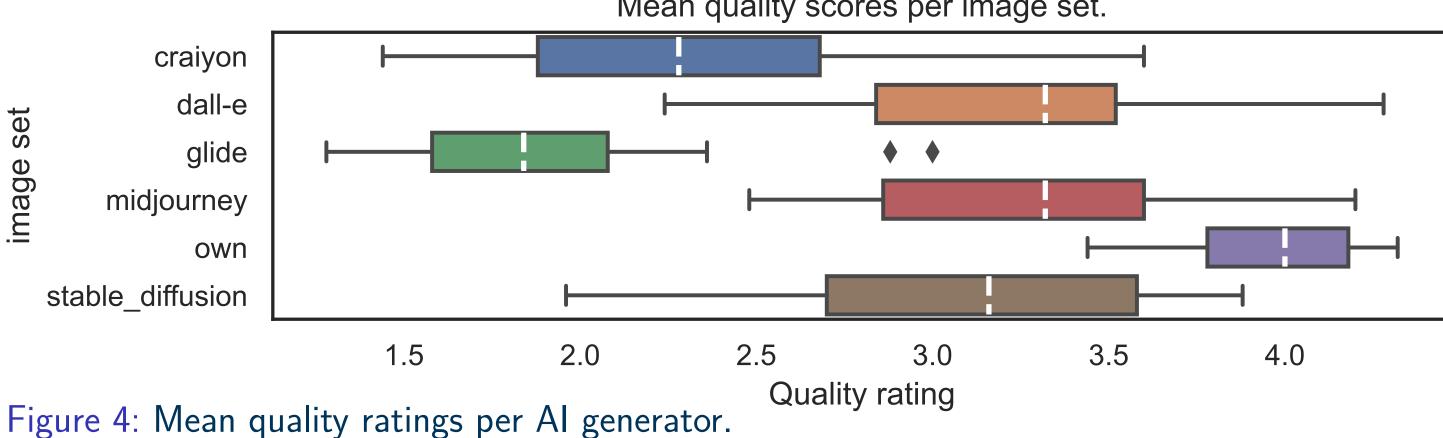


Figure 3: Mean appeal ratings per AI generator.

Conclusion

- Imited subjective evaluation for AI-generated images for different generators
- evaluation: AVT-AI-Image-Dataset appeal/ quality; crowdsourcing
- ► Glide and Craiyon: overall low appeal and quality

- ▶ best: "own" p20, DALL-E-2 p23; worst: Glide p27; see Fig. 2
- ▶ image quality models: best: MANIQA (0.44 PCC), BRISQUE (-0.39 PCC)
- ► appeal vs. quality:
 - \circ overall: 0.80 PCC, higher appeal \leftrightarrow higher quality
 - glide: 0.57 PCC; "own": 0.58 PCC
 - stable_diffusion: 0.62 PCC; dall-e: 0.63 PCC
 - midjourney: 0.74 PCC; craiyon: 0.77 PCC



Mean quality scores per image set.

- **Future Work**
- objective quality models: low performance for Al-generated images
- prediction models and features for Al-generated images
- ► larger datasets

► DALL-E-2 and Midjourney: similar high appeal/ quality to real photos

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► newer Al generators

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