## Supplementary material for It's All About *Your* Sketch: Democratising Sketch Control in Diffusion Models

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## **A. Additional Qualitative Results**

Fig. 1 delineates qualitative comparison of our method with pix2pix [1], CycleGAN [6], ControlNet [5], T2I-Adapter [2], and PITI [4]. Whereas, Fig. 2-7 shows additional results generated by our method.



Figure 1. Qualitative comparison with SoTAs. For ControlNet [5], T2I-Adapter [2], and PITI [4], we use the fixed prompt "a photo of [CLASS]", with [CLASS] replaced with corresponding class-labels of the input sketches. (*Best view when zoomed in.*)





Figure 3. Images generated by our method.











































Figure 4. Images generated by our method.













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Figure 5. Images generated by our method.





































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Figure 6. Images generated by our method.















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## **B.** Results on Out-of-distribution Sketches

Keeping the pre-trained diffusion model *frozen*, we fully leverage its generalisation potential. We posit that our design enables out-of-distribution generalisation. Fig. 8 shows a few sketches, absent in Sketchy [3].



Figure 8. Results on out-of-distribution sketches.

## References

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