

Figure 1: Illustration of CLIPScore and Directional CLIPScore. (a) CLIPScore evaluates the similarity between  $V_{img}^{CS}$  and  $V_{text}^{CS}$  from the coordinate origin, where the angle between the two vectors is bounded, resulting in a limited similarity value. (b) Directional CLIPScore gauges the similarity between  $V_{img}^{DCS}$  and  $V_{text}^{DCS}$  using the defined means of images  $C_{\mathcal{X}}$  and texts  $C_{\mathcal{T}}$  as the origin, the range of similarity is unrestricted.

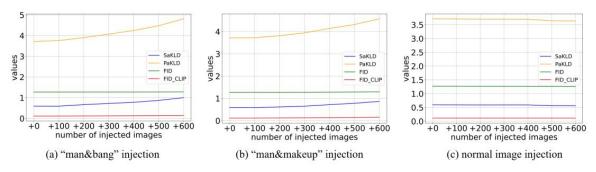


Figure 2: Validation of metrics by including correlated images. We set two distinct subsets of the FFHQ dataset (30,000 images each), and linearly injected correlated images hard to find into a subset (a)(b), and normal images (c). While SaKLD and PaKLD captures distribution changes, FID and  $FID_{CLIP}$  score remained stagnent.

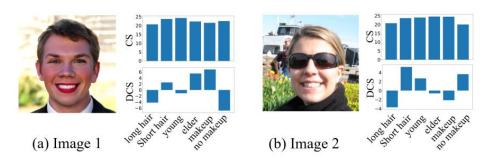


Figure 3: **Scores for opposite attributes.** We examined CS and DCS values for opposite attributes such as "long hair" and "short hair". In both images, Directional CLIPScore shows opposite directions for opposite attributes.



Figure 4: **Samples from ProjectedGAN.** While the FID of ProjectedGAN is comparable to major models, SaKLD and PaKLD of ProjectedGAN are significantly inferior because it fails to capture distributions of attributes, generating such a baby with beard.