

G8: Face Recognition from Caricature-Sketches

1) Can one go from a normal natural face image and generate a caricature face from it? What will be the salient feature points or control points? Can one use these control points to alter the face pose and perspective of the transformed caricature?

Yes, it is possible to generate a caricature from a normal natural face image using deep learning techniques, particularly Generative Adversarial Networks (GANs). Models such as CAFÉ-GAN, CariGANs, and StyleGAN-based methods have been shown to produce high-quality caricature transformations from real face photos. The key to generating a caricature lies in identifying salient facial landmarks or control points. These typically include: - Eyes (corners, centers) - Eyebrows - Nose (tip, bridge, nostrils) - Mouth (corners, top, bottom) - Jawline and chin - Cheekbones - Face contour. These feature points are extracted using tools like Dlib, OpenFace, or MediaPipe. Once extracted, these landmarks can be manipulated to exaggerate or warp the facial structure — enlarging eyes, elongating the chin, or widening the smile. Furthermore, these control points can also be used to alter the pose and perspective of the caricature. Techniques such as 3D Morphable Models (3DMM) or pose-guided GANs allow changes in the yaw, pitch, and roll angles of the face, offering control over how the caricature appears in different orientations.

2) From face transformed to caricature-variant, can one go back to the new face (full colour) seeded by a new perspective?

Yes, the process of transforming a caricature back into a realistic face — especially with a different pose or expression — is known as reverse translation or caricature-to-face mapping. This is achievable through bidirectional GAN architectures such as CycleGAN, DualGAN, or more specialized ones like CAFÉ-GAN. When a face is transformed into a caricature, some exaggerated features and distortions are introduced. To go back to a natural-looking face, the model needs to: - De-exaggerate the features - Restore texture, color, and fine details - Handle missing or ambiguous information. Using pose control mechanisms (e.g., pose vectors or latent direction manipulation), one can regenerate the original face with different angles or expressions. Thus, the original face can be reconstructed and re-synthesized under new viewpoints, enhancing its use for recognition purposes.

3) How does this approach improve face recognition through data augmentation?

The two-step transformation process — from face to caricature and then back to face (possibly under new perspectives) — offers a powerful form of data augmentation. Here's how it improves a face recognition system: 1. Diversity of Training Data: Multiple visual variants of the same identity are generated, increasing intra-class variation and helping the model learn more robust identity representations. 2. Handling Style Variations: Caricatures introduce extreme variations in style. Training on such inputs helps the model become more style-invariant, useful across different image modalities. 3. Pose and Expression Robustness: Seeding the reconstructed face with new poses or expressions helps recognize the same person across various orientations. 4. Improved Generalization: The pipeline forces the model to learn deeper, identity-preserving features, not just superficial textures. 5. Low-Resource Advantage: In cases with limited real data, such synthetic augmentation boosts model performance without needing large datasets. Overall, this approach improves recognition across domains, styles, poses, and conditions, making it practical for real-world applications.