

自动上妆进行了深入研究,取得了一定的成果,但在以后的工作中还有改进和优化的方面。

- 尝试对数据进行扩展,如光照、色彩对比度等,增强模型对人脸色彩生成的多样性。
- 尝试着重训练每个类别中有代表性的部分数据,尽可能地突出类别特点,同时兼顾其他数据,增强模型的泛化能力。

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包仁达(1994-),男,河北石家庄人,硕士,主要研究领域为生成对抗网络,图像语义分割.



黄少飞(1996-),女,硕士,主要研究领域为基于生成对抗网络的人脸编辑及应用.



庾涵(1992-),女,硕士,主要研究领域为图像语义分割,关键点检测.



孙瑶(1983-),男,博士,副研究员,主要研究领域为密码哈希函数分析,代数方程组求解算法研究,符号计算,计算机视觉.



朱德发(1994-),男,硕士,主要研究领域为图像及视频合成,迁移学习.



刘德(1985-),女,博士,副教授,CCF 专业会员,主要研究领域为深度学习,计算机视觉,多媒体分析.

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