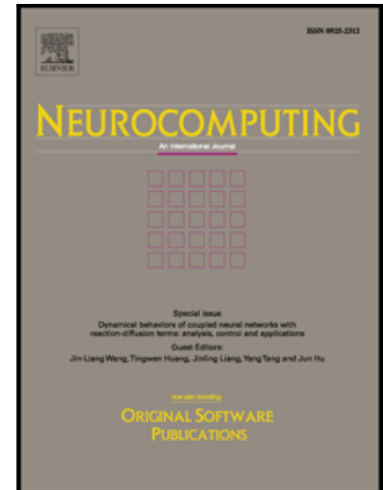


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# Cross-Heterogeneous-Database Age Estimation Through Correlation Representation Learning

Qing Tian<sup>1,2,3</sup>, Songcan Chen<sup>3\*</sup>

<sup>1</sup> Jiangsu Engineering Center of Network Monitoring, Nanjing University of Information Science and Technology, Nanjing 210044, P.R.China

<sup>2</sup> School of Computer and Software, Nanjing University of Information Science and Technology, Nanjing 210044, P.R.China

<sup>3</sup> College of Computer Science and Technology, Nanjing University of Aeronautics and Astronautics, Nanjing 211106, P.R.China

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## Abstract

Human age estimation is an important research topic and has found its applications in such scenarios as commodity recommendation and security monitoring. The design of existing estimators generally follows a same pipeline, i.e., an estimator is built from a given training dataset and then evaluated on a holdout testing set from the same dataset to display its effectiveness. In doing so, an implicit assumption is that both training and testing sets should share the same age distribution and feature representation, consequently-meaning that 1) once the age of a face image to be tested is outside the age range of training set, a mis-estimation is naturally inevitable; 2) an estimator built on a specific dataset usually cannot be directly applied to make evaluations on other datasets, because the dimensions and types of their feature representations are usually different (i.e., these datasets are heterogeneous). That is, existing methods can not be directly employed to perform cross-heterogeneous-dataset age estimation. To the best of our knowledge, the age distributions of existing aging datasets are usually not consistent but complementary to each other. Motivated by such a complementarity characteristic of different datasets in age distributions, we develop a so-called *correlation component manifold space learning* (CCMSL) to first learn a common feature space by capturing the correlations between the heterogeneous databases, and then in the resulting space establish a single age estimator across such heterogeneous datasets through *correlation representation learning* (CRL). As a result, not only can the age-distribution-incompleteness of individual aging datasets be compensated, but also the discriminating ability of the estimator be reinforced. Finally, experimental results demonstrate the superiority of the proposed methods.

**Keywords:** Age Estimation, Cross-Heterogeneous-Database, Correlation Component Manifold Space Learning, Correlation Representation Learning

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## 1. Introduction

Human age estimation is an important research topic in computer vision community and has found its wide applications in such scenarios as recommendation systems [27], [32], security access control [18], [40], biometrics [38], [31] and entertainment [6], [9], etc. To perform human facial appearance age estimation, a variety of methods have been developed. Generally, these methods can be grouped into three categories: classification-based (e.g., [23], [13], [37], [2], [33], [1], [8], [16]), regression-based (e.g., [24], [12], [28], [41], [42], [14], [11], [3], [29], [26], [25], [4], [15]), and their hybrid methods (e.g., [18], [22]). When treating each age as a separate class, we can perform age estimation using existing classification frameworks. According to this principle, artificial neural networks (ANN) [23], conditional probability neural networks (CPNN) [13], Gaussian mixture models [37], and extreme learning machines (ELM) [33] have been used to classify human age. Recently, Alnajjar et al. [1] proposed an expression-insensitive age estimation method. Dibeklioglu et al. [8] proposed to estimate human age by taking into account the facial dynamics together with static appearance information. Actually, age estimation is more of a regression problem rather than an ordinary classification problem due to its characteristics of continuity and monotonicity. Based on this fact, quadratic function [24], [11], multiple linear regression [12],  $\xi$ -SVR [28], SDP regressors [41, 42], aging pattern subspace

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Corresponding author: s.chen@nuaa.edu.cn

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