

The proposed age estimation algorithm ~~realises~~realizes hierarchical approach (Fig. 10).

~~First of all, the~~ input fragments are divided ~~into~~for three age groups: ~~less~~smaller than 18 years old, ~~from 18—45 years old,~~ and ~~more~~bigger than 45 years old. ~~Second, Afterwards~~ the results of this ~~in the~~ first step~~age~~ are ~~further~~more subdivided ~~in~~to seven ~~smaller~~newer groups, ~~with each limited~~ing to ~~a~~one single decade. ~~This reduces~~us the ~~original~~problem of multiclass classification ~~problem~~is therefore reduced to a set of binary “one-against-all”-classifiers (~~;~~ BCs). ~~Each~~These classifiers ~~calculate: then~~ ranks ~~the~~of images ~~each based on~~of the ~~associated~~analyzed class, ~~and-~~ ~~The final~~total decisions ~~are~~is ~~obtained~~then by the ~~analyz~~ing these ~~previously received~~ rank histograms of ranks.

~~A two-level schemes of~~ ~~These binary classifier~~ BCs are constructed ~~using a two-level~~ approach. ~~After ion is applied first~~with the transitioning to ~~an~~ adaptive feature space, ~~as equal~~ ~~to this~~ described earlier, ~~the images are-~~ classified using ~~and~~ support vector machines ~~s~~ classification with ~~radial basis function~~ RBF kernels.

~~The i~~ Input fragments ~~are~~were preprocessed ~~for their luminance characteristics~~ to align and ~~to~~ transform ~~their luminance characteristics~~ ~~them~~ to a uniform ~~al~~ scale. ~~This p~~ Preprocessing step includes color-space transformation and scaling, both operations similar to those used in ~~the~~that of a gender recognition algorithm. Features, ~~are~~ calculated for each colour component ~~and-~~ ~~are~~ combined to form a uniform feature vector.

Training and testing require a sufficiently ~~huge~~ large ~~enough~~ coloring image database. ~~Here,~~ ~~We combined~~used the state-of-the-art ~~image databases~~ MORPH and FG-NET image databases with our own image database, gathered from many-different sources ~~and,~~ ~~which~~ comprising ~~ed~~ of 10,500 face images. ~~The f~~ Faces ~~in~~ the images were detected automatically by the AdaBoost face detection algorithms.

Comment [A1]: “First of all” is not incorrect, but it is unnecessarily verbose. In addition, simply using “First” makes the presentation more uniform if you go on to discuss “second” or “third” points, for example.

Comment [A2]: The words “smaller” and “bigger” are typically used to, say, discuss the size of objects. If you want to compare numbers, “less than” and “more than” (or “greater than”) are more appropriate.

Comment [A3]: Here, “histogram of ranks” is not wrong, but feels awkward and unnatural. Often, we can take a phrase such as “A of B” and turn it around like this to give just “B A,” eliminating the “of.” (Note that now, “rank” is singular.)

Comment [A4]: Here, “to align and to transform” is not wrong, but sounds awkward. When using an “A and B” construction like this, we typically keep the A and B parts as small as possible, placing any common parts before the main construction.

Comment [A5]: In everyday usage, it may seem natural to talk about a vector of features as a featured vector, but the standard term in the field is “feature vector.”

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A total number of ~~seven thousand~~ 7000 images were used to train and test the first stage of the for age classification algorithm ~~training and testing on the first stage~~. ~~Three~~ 3 binary classifier BCs were ~~created, each made with~~ utilizing 144 adaptive features ~~each of~~.

The first-stage cClassification results ~~showed on the first stage are~~: 82-% accuracy for young faces~~age~~, 58-% accuracy for middle-~~aged faces~~, and 92-% accuracy for ~~elderly faces~~ senior age. The overall aAge classification ~~accuracy for rate in the~~ three age categories was ~~division problem~~ – 77.3-%.

The second-stage BC~~Binary classifiers of the second stage~~ were constructed ~~in the same way as for~~ equal to the first stage (~~described above~~). Fig. 11 shows a~~A~~ visual example of age estimation by the ~~first stage of the~~ proposed algorithm ~~on its first stage is presented in figs. 11~~.