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Information technology — Biometric sample quality —

Part 4: **Finger image data**

Technologies de l'information — Qualité d'échantillon biométrique — Partie 4: Données d'image de doigt





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Foreword

ISO (the International Organization for Standardization) and IEC (the International Electrotechnical Commission) form the specialized system for worldwide standardization. National bodies that are members of ISO or IEC participate in the development of International Standards through technical committees established by the respective organization to deal with particular fields of technical activity. ISO and IEC technical committees collaborate in fields of mutual interest. Other international organizations, governmental and non-governmental, in liaison with ISO and IEC, also take part in the work. In the field of information technology, ISO and IEC have established a joint technical committee, ISO/IEC ITC 1.

The procedures used to develop this document and those intended for its further maintenance are described in the ISO/IEC Directives, Part 1. In particular the different approval criteria needed for the different types of ISO documents should be noted. This document was drafted in accordance with the editorial rules of the ISO/IEC Directives, Part 2 (see www.iso.org/directives).

Attention is drawn to the possibility that some of the elements of this document may be the subject of patent rights. ISO shall not be held responsible for identifying any or all such patent rights. Details of any patent rights identified during the development of the document will be in the Introduction and/or on the ISO list of patent declarations received (see www.iso.org/patents).

Any trade name used in this document is information given for the convenience of users and does not constitute an endorsement.

For an explanation on the voluntary nature of standards, the meaning of ISO specific terms and expressions related to conformity assessment, as well as information about ISO's adherence to the World Trade Organization (WTO) principles in the Technical Barriers to Trade (TBT) see the following URL: www.iso.org/iso/foreword.html.

This document was prepared by Joint Technical Committee ISO/IEC JTC 1, *Information technology*, Subcommittee SC 37, *Biometrics*.

This first edition cancels and replaces ISO/IEC/TR 29794-4:2010, which has been technically revised to become an International Standard.

A list of all parts in the ISO 29794 series can be found on the ISO website.

Introduction

This document specifies finger image quality metrics. A reference implementation of the normative metrics is available at https://github.com/usnistgov/NFIQ2.

The quality of finger image data is defined to be the degree to which the finger image data fulfils specified requirements for the targeted application. Thus, the quality information is useful in many applications. ISO/IEC 19784-1 allocates a quality field and specifies the allowable range for the scores, with a recommendation that the score be divided into four categories with a qualitative interpretation for each category. Image quality fields are also provided in the fingerprint data interchange formats standardized in ISO/IEC 19794-2, ISO/IEC 19794-3, ISO/IEC 19794-4, and ISO/IEC 19794-8. This document defines a standard way to calculate the finger image quality score that facilitates the interpretation and interchange of the finger image quality scores.

Information technology — Biometric sample quality —

Part 4:

Finger image data

1 Scope

This document establishes

- terms and definitions for quantifying finger image quality,
- methods used to quantify the quality of finger images, and
- standardized encoding of finger image quality,

for finger images at 196,85 px/cm spatial sampling rate scanned or captured using optical sensors with capture dimension (width, height) of at least 1,27 cm × 1,651 cm.

2 Normative references

The following documents are referred to in the text in such a way that some or all of their content constitutes requirements of this document. For dated references, only the edition cited applies. For undated references, the latest edition of the referenced document (including any amendments) applies.

ISO/IEC 2382-37, Information technology — Vocabulary — Part 37: Biometrics

ISO/IEC 19794-1:2011, Information technology — Biometric data interchange formats — Part 1: Framework

ISO/IEC 29794-1, Information technology — Biometric sample quality — Part 1: Framework

3 Terms, definitions, symbols and abbreviated terms

3.1 Terms and definitions

For the purposes of this document, the terms and definitions given in ISO/IEC 2382-37, ISO/IEC 29794-1 and the following apply.

ISO and IEC maintain terminological databases for use in standardization at the following addresses:

- IEC Electropedia: available at http://www.electropedia.org/
- ISO Online browsing platform: available at http://www.iso.org/obp

3.1.1

foreground region

set of all pixels of a finger image that form valid finger image patterns

Note 1 to entry: The most evident structural characteristic of a valid finger image is a pattern of interleaved ridges and valleys.

3.1.2

local region

block of $m \times n$ pixels of the foreground of a finger image, where m and n are smaller than or equal to the width and the height of the finger image

3.1.3

finger image quality assessment algorithm

algorithm that reports a quality score for a given finger image

3.1.4

metric

quantification of a covariate using a prescribed method

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covariate

variable or parameter that either directly, or when interacting with other covariates, affects fingerprint recognition accuracy

3.2 Symbols and abbreviated terms

DFT Discrete Fourier Transform

- I matrix of grey-level intensity values corresponding to the pixels of an image
- *S* ridge valley signature of a local region V
- *V* matrix of grey-level intensity values corresponding to the pixels of a local region

4 Conformance

A finger image quality assessment algorithm conforms to this document if it conforms to the normative requirements of <u>Clause 5</u>.

A finger image quality record shall conform to this document if its structure and data values conform to the formatting requirements of <u>Clause 6</u> (finger image quality data record) and its quality values are computed using the methods specified in <u>5.2</u>, <u>5.3</u> and <u>5.4</u>.

Conformance to normative requirements of <u>Clause 6</u> fulfils Level 1 and Level 2 conformance as specified in ISO/IEC 19794-1:2011, Annex A. Conformance to normative requirements of <u>5.2</u> and <u>5.4</u> is Level 3 conformance as specified in ISO/IEC 19794-1:2011, Annex A.

5 Finger image quality metrics

5.1 Overview

5.1.1 General

<u>Clause 5</u> establishes metrics for predicting the utility of a finger image (5.2 and 5.3). Image quality metrics from a single image are useful to ensure the acquired image is suitable for recognition.

A complete finger image quality analysis shall examine both the local and global structures of the finger image. Fingerprint local structure constitutes the main texture-like pattern of ridges and valleys within a local region while valid global structure puts the ridges and valleys into a smooth flow for the entire fingerprint. The quality of a finger image is determined by both its local and global structures. Clause 5 describes the features and characteristics of finger images at both local and global structures that are to be used for quantifying finger image quality.

For applying the algorithms as described in <u>5.2</u> and <u>5.3</u>, the finger image shall have a spatial sampling rate of 196,85 pixels per centimetre (500 pixels per inch).

5.1.2 Constituent of local quality metrics

A finger image is partitioned into local regions such that each local region contains sufficient ridge-valley information, preferably having at least 2 clear ridges, while not overly constraining the high curvature ridges. For images with a spatial sampling rate of 196,85 pixel per centimetre (500 pixel per inch), the ridge separation usually varies between 8 pixels to 12 pixels^[1]. A ridge separation comprises a ridge and a valley. In order to cover two clear ridges, the local region size has to be greater than 24 pixels in both width and height. The size for each local region shall be 32×32 pixels, which is sufficient to cover 2 clear ridges. Instead of Cartesian coordinate, curvilinear coordinate along the ridge can also be used.

5.1.3 Constituent of global quality metrics

A global quality metric should be computed over the whole image and assess the utility of fingerprint characteristics in the image.

5.1.4 Image preprocessing

5.1.4.1 Description

A segmentation process follows where each local region is labelled as background or foreground. There are several segmentation approaches, such as using the average magnitude of the pixel-intensity gradient in each local region[1].

This document does not prescribe segmentation methods, but notes that performing segmentation influences the computed scores. Constant or near constant areas of the input image shall be removed according to <u>5.1.4.2</u> prior to computing quality using the metrics specified in <u>5.2</u> and <u>5.3</u>.

5.1.4.2 Removal of near constant white lines in image

Prior to computing features, fingerprint images are cropped to remove white pixels on the margins. Starting from the outer margins, rows and columns with average pixel intensity above 250 are removed.

Pixel intensities take values [0, 255] for an 8-bit gray scale image. As a first approximation of the region of interest, image columns and rows which are near constant white background are removed. Using the algorithm specified below, a fixed threshold is set for gray scale pixel intensity of T_{μ} = 250 to obtain the image without near constant areas.

The algorithm is specified as:

- a) For each row R_i in I, starting from the top
 - 1) Compute the row arithmetic mean μ_{row}
 - 2) On the first occurrence where $\mu_{\text{row}} \leq T_u$ set $idx_t = i$
 - 3) On the last occurrence where $\mu_{\text{row}} \le T_u$ set $idx_b = i$
- b) For each column C_i in I, starting from the left
 - 1) Compute the column arithmetic mean μ_{col}
 - 2) On the first occurrence where $\mu_{col} \le T_u$ set $idx_l = i$
 - 3) On the last occurrence where $\mu_{col} \le T_u$ set $idx_r = i$
- c) Extract the region of interest as $\hat{I} = I$.roi $(idx_l, idx_t, idx_t, idx_t)$

5.1.4.3 Foreground segmentation based on local standard deviation

For quality features which require a foreground mask to indicate regions containing the fingerprint an algorithm using local standard deviation is adopted.

The algorithm is specified as:

- a) Normalize I to zero mean and unit standard deviation to produce \hat{I}
- b) For each local region V in \hat{I}
 - 1) Compute the standard deviation of V as σ_V
 - 2) Mark the corresponding local region in I_{mask} as foreground if $\sigma_V > 0.1$

5.1.4.4 Computing the dominant ridge flow orientation for a local region from pixel-intensity gradients

The dominant ridge flow orientation is determined by computing the pixel-intensity gradient information and then determining the orientation of the principal variation axis.

The numerical gradient of the local region is determined using finite central difference for all interior pixels in x-direction and y-direction

$$f_x = \frac{I(x+1,y) - I(x-1,y)}{2} \tag{1}$$

$$f_{y} = \frac{I(x, y+1) - I(x, y-1)}{2}$$
 (2)

With f_x and f_y , the dominant ridge flow orientation, *angle* (V), is determined analytically using the sine and cosine doubled angle determined from the arithmetic means of the pixel-intensity gradient covariances.

$$a = \overline{f_{\chi}^2} \tag{3}$$

$$b = \overline{f_y^2} \tag{4}$$

$$c = \overline{f_x f_y} \tag{5}$$

$$\mathbf{C} = \begin{bmatrix} a & c \\ c & b \end{bmatrix} \tag{6}$$

$$d = \sqrt{c^2 + \left(a - b\right)^2} + \epsilon \tag{7}$$

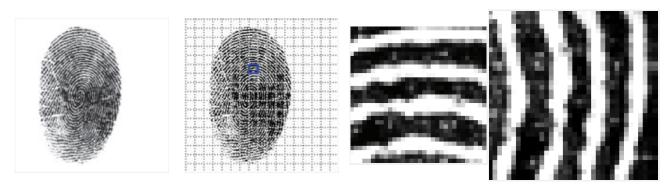
$$\sin\theta = \frac{c}{d} \tag{8}$$

$$\cos\theta = \frac{a-b}{d} \tag{9}$$

$$angle(\mathbf{V}) = \frac{1}{2} \tan^{-1} \frac{\sin \theta}{\cos \theta} \tag{10}$$

5.1.5 Image examples

For algorithms operating in a block-wise manner the input image is subdivided into local regions according to the overlay grid shown in Figure 1 b). The local region V(8,5) is used as example in local processing and is marked up using a bold line. Figure 1 c) shows an enlarged view of V(8,5) and Figure 1 d) shows V(8,5) rotated according to its dominant ridge orientation computed using Formula (10).



- a) Input finger image
- b) division into local regions
- c) enlarged view of *V*(8,5)
- d) V(8,5) rotated according to its dominant ridge orientation as determined using Formula (10)

Figure 1 — Input image used — Examples of the processing of quality

5.2 Normative contributive quality metrics

5.2.1 General

5.2 specifies normative contributive finger image quality assessment algorithms.

5.2.2 Orientation certainty level

5.2.2.1 Description

The orientation certainty level (OCL)[3] of a local region is a measure of the consistency of the orientations of the ridges and valleys contained within this local region. The feature computes local quality and operates in a block-wise manner.

The finger image within a 32×32 pixels local region [as shown in Figure 1 c)] generally consists of dark ridge lines separated by white valley lines along the same orientation. The consistent ridge orientation and the appropriate ridge and valley structure are distinguishable local characteristics of the fingerprint local region.

The pixel-intensity gradient (dx, dy) at a pixel describes the direction of the maximum pixel-intensity change and its strength. By performing Principal Component Analysis on the pixel-intensity gradients in a local region, an orthogonal basis for the local region can be formed by finding its eigenvalues and eigenvectors. The resultant first principal component contains the largest variance contributed by the maximum total gradient change in the direction orthogonal to ridge orientation. The direction is given by the first eigenvector and the value of the variance corresponds to the first eigenvalue, λ_{max} . On the other hand, the resultant second principal component has the minimum change of gradient in the direction of ridge flow which corresponds to the second eigenvalue, λ_{min} . The ratio between the two eigenvalues thus gives an indication of how strong the energy is concentrated along the dominant

direction with two vectors pointing to the normal and tangential direction of the average ridge flow respectively.

5.2.2.2 Computing the eigenvalues and local orientation certainty

From the covariance matrix C [Formula (6)] the eigenvalues λ_{min} and λ_{max} are computed as

$$\lambda_{\min} = \frac{a + b - \sqrt{(a - b)^2 + 4c^2}}{2} \tag{11}$$

$$\lambda_{\text{max}} = \frac{a + b + \sqrt{(a - b)^2 + 4c^2}}{2} \tag{12}$$

which yields a local orientation certainty level

$$\mathbf{Q}_{\text{OCL}}^{local} = \begin{cases} 1 - \frac{\lambda_{\min}}{\lambda_{\max}}, & \text{if } \lambda_{\max} > 0 \\ 0, & \text{otherwise} \end{cases}$$
 (13)

which is a ratio in the interval [0,1] where 1 is highest certainty level and 0 is lowest.

NOTE The orientation certainty level fails to predict match-ability when some marks or residual exist in the samples that have strong orientation strength, such as those exhibited by latent prints left by the previous user.

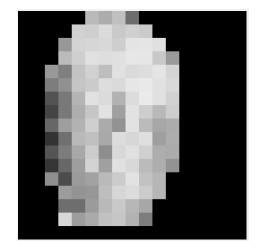
5.2.2.3 OCL algorithm

For each local region **V** in **I**:

- a) compute the pixel-intensity gradient of V with centered differences method [Formulae (1), (2)];
- b) compute the covariance matrix *C* [Formula (6)];
- c) compute the eigenvalues of \boldsymbol{c} to obtain $\boldsymbol{Q}_{\mathrm{OCL}}^{local}$ [Formulae (11), (12), (13)].

Figure 2 visualizes the processing steps.





a) Current local region with the ratio between eigenvalues marked as ellipse

b) Local quality scores $m{Q}_{ ext{OCL}}^{local}$ for example fingerprint image

Figure 2 — Processing steps of orientation certainty level quality algorithm

5.2.3 Local clarity score

5.2.3.1 Description

Good quality fingerprints exhibit clear ridge-valley structure. Thus, the local clarity score (LCS) [4], which is the measure of the ridge-valley structure clarity, is a useful indicator of the quality of a fingerprint. The feature computes local quality and operates in a block-wise manner.

To perform ridge-valley structure analysis, the foreground of the finger image is quantised into local regions of size 32×32 pixels^[3]. Inside each local region, an orientation line, which is perpendicular to the ridge direction, is computed. At the centre of the local region along the ridge direction, a local region of size 32×16 pixels shall be extracted and transformed to a vertically aligned local region.

On S, the local region average profile, calculated in 5.2.3.4, a linear regression (or least square fitting) is applied to determine the Determine Threshold (DT) which is a line positioned at the centre of the local region V, and is used to segment the local region into the ridge or valley region. Regions with grey level intensity lower than DT are classified as ridges. Otherwise, they are classified as valleys.

Since good finger images cannot have ridges that are too close or too far apart, the nominal ridge and valley thickness can be used as a measure of the quality of the finger image captured. Similarly, ridges that are unreasonably thick or thin indicate that the finger image may not be captured properly, such as pressing too hard or too soft, or the image is a residual sample. Thus, the finger image quality can be determined by comparing the ridge and valley thickness to each of their nominal range of values. Any value out of the nominal range may imply a bad quality ridge pattern. To normalize the range of the thickness values, a pre-set maximum thickness is used. The maximum ridge or valley thickness (W_{max}) for a good finger image is estimated at 20 pixels for a 196,85 pixel per centimetre (500 pixel per inch). The pre-set value of 20 pixel for a 196,85 pixel per centimetre (500 pixel per inch) scanner spatial sampling rate is obtained from the median of the typical ridge separation of 8 to 12 pixels [1], and assuming that any ridge separation will not exceed twice of the median value. This will ensure that the pre-set value is indeed the maximum to limit the value of the normalized ridge and valley thickness between 0 and 1. The ridge thickness (W_r) and valley thickness (W_v) are then normalized with respect to the maximum thickness.

With the ridge and valley separated as above, a clarity test can be performed in each segmented rectangular 2-D region.

For local regions with good clarity, the pixel-intensity distribution of ridges and the pixel-intensity distribution of the valleys have a very small overlapping area and thus Q_{LCS}^{local} is high. The following factors affect the size of the total overlapping area:

- a) noise on ridge and valley;
- b) water patches on the image due to wet fingers;
- c) incorrect orientation angle due to the effect of directional noise;
- d) scar across the ridge pattern;
- e) highly curved ridges;
- f) ridge endings, bifurcations, delta and core points;
- g) incipient ridges, sweat pores and dots.

Factors a) to c) are physical noise found in the image. Factors d) to g) are actual physical characteristics of the fingerprint.

5.2.3.2 Computing the ridge valley signature of a local region

Given the local region *V* the ridge valley signature *S* is obtained by

$$S(x) = \frac{\sum_{y=1}^{16} V(x, y)}{16}$$
 (14)

where V(x, y) is the grey level at point (x, y); x is the index along x-axis.

5.2.3.3 Determining the proportion of misclassified pixels

Formulae (15) and (16) specify the calculation of α and which are the proportion of pixels misclassified respectively as valley or ridge. v_B is the number of pixels in valley region with intensity lower than DT and v_T is the total number of pixels in valley region. r_B is the number of pixels in the ridge region with intensity higher than DT and r_T is the total number of pixels in the ridge region.

$$\alpha = \frac{v_B}{v_T} \tag{15}$$

$$\beta = \frac{r_B}{r_T} \tag{16}$$

5.2.3.4 Determining the normalized ridge and valley width

The normalized valley width \overline{W}_v and the normalized ridge width \overline{W}_r are determined

$$\overline{W}_{v} = \frac{W_{v}}{\left(\frac{S}{125}\right)} W^{\text{max}}$$

$$\overline{W}_{r} = \frac{W_{r}}{\left(\frac{S}{125}\right)} W^{\text{max}}$$
(18)

$$\overline{W}_r = \frac{W_r}{\left(\frac{S}{125}\right)W^{\text{max}}} \tag{18}$$

where

S is the scanner spatial sampling rate in dpi;

Mmax is the estimated ridge or valley width for an image with 49,21 pixel per centimetre (125 pixel per inch) spatial sampling rate;

 W_v and W_r are the observed valley and ridge widths.

According to Reference [1], $W^{\text{max}} = 5$ is reasonable for 49,21 pixel per centimetre (125 pixel per inch) spatial sampling rate. By extension, the denominator in Formula (17) and the denominator in Formula (18) shall be 20 for a spatial sampling rate of 196,85 pixels per centimetre (500 pixels per inch).

5.2.3.5 Computing the local clarity score

The local quality score Q_{LCS}^{local} is the constrained average value of α and β with a range between 0 and 1.

$$Q_{\text{LCS}}^{local} = \begin{cases} 1 - \frac{\alpha + \beta}{2}, & \text{if } \left(W_v^{nmin} < \overline{W_v} < W_v^{nmax}\right), \left(W_r^{nmin} < \overline{W_r} < W_r^{nmax}\right) \\ 0, & \text{otherwise} \end{cases}$$

$$(19)$$

where

 W_r^{nmin} and W_v^{nmin} are the minimum values for the normalized ridge and valley width;

 W_v^{nmax} and W_v^{nmax} are the maximum values for the normalized ridge and valley width.

$$W_r^{nmin} = \frac{3}{W_r} \tag{20}$$

$$W_r^{nmax} = \frac{10}{\overline{W_r}} \tag{21}$$

$$W_{v}^{nmin} = \frac{2}{W_{v}} \tag{22}$$

$$W_{v}^{nmin} = \frac{10}{W_{v}} \tag{23}$$

NOTE Particular regions inherent in a fingerprint will negatively affect Q_{LCS}^{local} . For example, ridge endings and bifurcations or areas with high curvature such as those commonly found in core and delta points.

5.2.3.6 LCS algorithm

For each local region V in I:

- a) rotate V such that dominant ridge flow is perpendicular to x-axis;
- b) crop rotated V such that no invalid regions are included;
- c) with V obtain the ridge-valley signature S(5.2.3.2);
- d) determine DT using linear regression on S;
- e) for each element S(x), set threshold T(x) of x being ridge or valley based on DT;
- f) classify columns in V as ridge (1) or valley (0) with $P(x) = \begin{cases} 1, & \text{if } S(x) < T(x) \\ 0, & \text{otherwise} \end{cases}$;
- g) determine ridge-valley transition vector **C** from **P**;
- h) compute the vector \boldsymbol{W} containing ridge and valley widths from \boldsymbol{C} ;
- i) determine normalized ridge width and valley width \overline{W}_r and \overline{W}_v (5.2.3.4);
- j) determine the proportion of misclassified pixels α and β (5.2.3.3);
- k) compute the local quality score Q_{LCS}^{local} (5.2.3.5).

Figure 3 visualizes the processing steps.

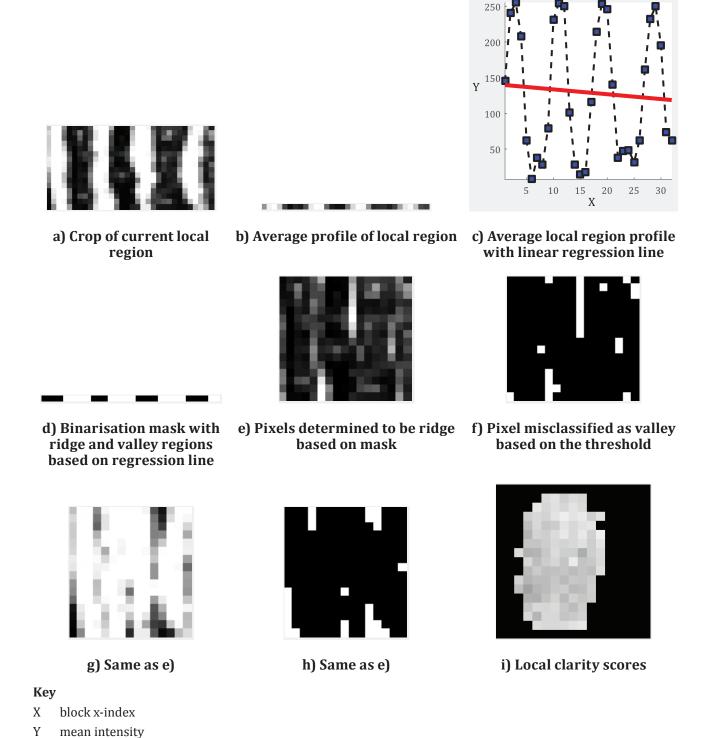


Figure 3 — Processing steps of local clarity score algorithm

5.2.4 Frequency domain analysis (FDA) score

5.2.4.1 Description

Frequency domain analysis (FDA) computes local quality and operates in a block-wise manner. A one-dimensional signature of the ridge-valley structure is extracted and the Discrete Fourier Transform (DFT) is computed on the signature to determine the frequency of the sinusoid following the ridge-valley structure [5].

The ridge-valley signature of a high quality sample is a periodic signal, which can be approximated either by a square wave or a sinusoidal wave. In the frequency domain, an ideal square wave should exhibit a dominant frequency with sideband frequency components (sinc function). A sinusoidal wave consists of one dominant frequency and minimum components at other non-dominant frequencies.

For each local region, a signature perpendicular to the dominant ridge flow orientation is computed.

The FDA described in 5.2.4 computes the one-dimensional signatures by performing averaging along the ridge flow direction. The averaging process filters out noise along the ridge and valley flow and provides a modelling of a smooth changing signal in a direction perpendicular to ridge flow.

5.2.4.2 Computing the local FDA quality score

The local quality score is computed by using Formula (24):

$$\mathbf{Q}_{FDA}^{local} = \begin{cases} 1, & if \quad F_{\text{max}} = \mathbf{A}_{1} \quad \text{or} \quad F_{\text{max}} = \mathbf{A}_{|A|} \\ \frac{\mathbf{A}_{F_{\text{max}}} + C\left(\mathbf{A}_{F_{\text{max}-1}} + \mathbf{A}_{F_{\text{max}+1}}\right)}{\sum_{F=1}^{|A|/2} A_{F}}, & \text{otherwise} \end{cases}$$
(24)

where

0,3 is the attenuation parameter *C*;

A is the amplitude at frequency index *x*.

The value of $\mathbf{Q}_{\mathrm{FDA}}^{local}$ is set to 1 when the maximum frequency F_{max} amplitude occurs at index $F_{\mathrm{max}} = \mathbf{A}_1$ or $F_{\mathrm{max}} = \mathbf{A}_{|\mathbf{A}|}$.

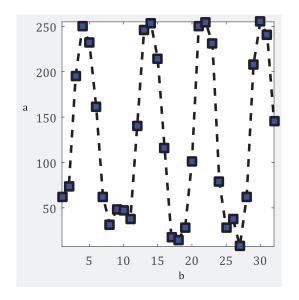
5.2.4.3 FDA algorithm

For each local region V in I:

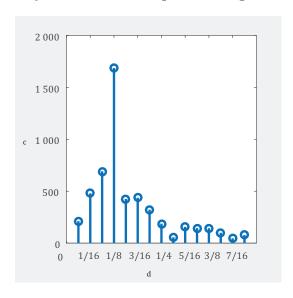
- a) pad *V* with 2 pixel border;
- b) rotate V with nearest neighbour interpolation such that dominant ridge flow is perpendicular to x-axis;
- c) crop **V** such that no invalid regions are included;
- d) with V obtain the ridge-valley signature S(5.2.3.2);
- e) compute the DFT of **S** to obtain the magnitude representation **A**;
- f) discard the first component of *A*;
- g) determine F_{max} as the index with the largest magnitude in A;
- h) compute $\mathbf{Q}_{\text{FDA}}^{local}$ of \mathbf{V} using \mathbf{A} and F_{max} (5.2.4.2).

Figure 4 visualizes the processing steps.

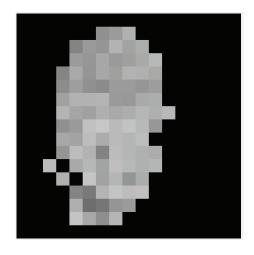




a) Central area of input local region



b) Ridge-valley profile



c) DFT of ridge-valley profile

- a Mean intensity.
- b Block x-index.
- c Magnitude.
- d Cycle/pixel.

d) Map of Q_{FDA}^{local}

Figure 4 — Processing steps of FDA quality algorithm

5.2.5 Ridge valley uniformity

5.2.5.1 Feature description

Ridge valley uniformity (RVU) is a measure of the consistency of the ridge and valley widths[3]. The expectation for finger image with clear ridge and valley separation is that the ratio between ridge and valley widths remains fairly constant throughout the finger image.

The ratio of ridge thickness to valley thickness should be constant and close to 1 throughout the whole image for a good quality finger image. The feature computes local quality and operates in a block-wise manner.

5.2.5.2 RVU algorithm

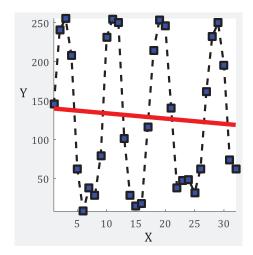
For each local region *V* in *I*:

- a) determine dominant ridgeflow orientation *angle* (V) of V;
- b) rotate V such that angle(V) is perpendicular to x-axis;
- c) crop **V** such that no invalid regions are included;
- d) with V obtain the ridge-valley signature S(5.2.3.2);
- e) determine DT using linear regression on *S*;
- f) for each S(x) compute threshold $T(x) = x \times DT(1) + DT(0)$;
- g) binarize S using T;
- h) classify ridge and valley in **S** as $P(x) = \begin{cases} 1, & \text{if } S(x-1) < T(x) \\ 0, & \text{otherwise} \end{cases}$;
- i) compute ridge-valley transition vector as $C(x) = \{ 1, \text{ if } P(x-1) \neq P(x) \}$;
- j) Drop first and last transition from **S** using **C** to remove incomplete ridges or valleys and obtain **S**';
- k) Compute Q_{FDA}^{local} as the ratio between widths of ridge and valleys in S'.

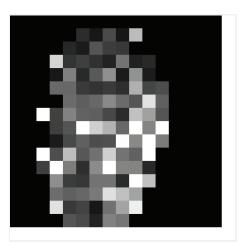
Figure 5 visualizes the processing steps.



a) Crop of current local region



b) Average profile of local region



c) Average profile with regression line

d) Local quality score as the standard deviation of local ridge to valley ratios

Key

X block x-index

Y mean intensity

Figure 5 — Processing steps of ridge valley uniformity quality algorithm

NOTE The ridge valley uniformity quality feature is spatial sampling rate dependent. The given defaults assume 196,85 pixel per centimetre (500 pixel per inch).

5.2.6 Orientation flow

5.2.6.1 Description

Orientation flow $(OFL)^{[\underline{4}]}$ is a measure of ridge flow continuity which is based on the absolute orientation difference between a local region and its 8-neighborhood of local regions.

Orientation flow is a good indicator to describe the quality of a good fingerprint pattern because, in general, the flow of the ridge direction changes gradually, except in an area with a delta or a core. The feature computes local quality and operates in a block-wise manner.

5.2.6.2 Local region-wise absolute orientation difference

The ridge flow is determined as a measure of the absolute difference between a local region and its neighboring local regions. The absolute difference D(i, j) for local region V(i, j) is computed using the dominant ridge flow orientations of this local region and of its neighbors

$$\mathbf{D}(i,j) = \frac{\sum_{m=-1}^{1} \sum_{n=-1}^{1} \left| angle(\mathbf{V}(i,j)) - angle(\mathbf{V}(i-m,j-n)) \right|}{8}$$
(25)

5.2.6.3 Local orientation flow quality score

The local orientation quality score Q_{OFL}^{local} for the local region orientation difference D(i,j) is

$$\mathbf{Q}_{OFL}^{local} = \begin{cases}
\frac{\mathbf{D}(i,j) - \theta_{\min}}{90^{\circ} - \theta_{\min}}, & \text{if } \mathbf{D}(i,j) > \theta_{\min} \\
0, & \text{otherwise}
\end{cases}$$
(26)

where θ_{min} = 4 is the threshold for minimum angle difference to consider.

5.2.6.4 OFL algorithm

- a) Determine the dominant ridge flow orientation angle(V) of local region V in I.
- For each local region *V* in *I*:
 - 1) compute the absolute orientation difference D(i, j) using angle (V) (5.2.6.2);
 - 2) compute the local orientation quality score $Q_{
 m OFL}^{local}$ (5.2.6.3).

Figure 6 visualizes the processing steps.

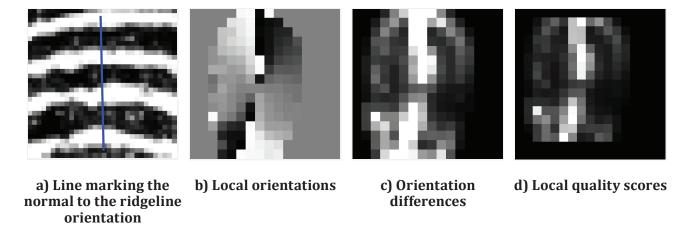


Figure 6 — Processing steps of orientation flow quality algorithm

5.2.7 MU

5.2.7.1 Description

The MU quality feature is the arithmetic mean of the pixel intensities of all pixels in the input image. The feature computes global quality.

5.2.7.2 MU algorithm

Compute $Q_{\rm MU}$ as the arithmetic mean of pixel intensities in *I*.

5.2.8 **MMB**

5.2.8.1 Description

The MMB quality feature is the arithmetic mean of per local region computed arithmetic mean in the gray scale input image. The feature computes local quality and operates in a block-wise manner.

5.2.8.2 MMB algorithm

- a) For each local region V in I
 - 1) compute the arithmetic mean of the pixel intensities in V as Q_{MMB}^{local} .
- b) Compute $Q_{\rm MMB}$ as the arithmetic mean of set of $\boldsymbol{Q}_{\rm MMB}^{local}$.

5.2.9 Minutiae count in finger image

5.2.9.1 Description

The FingerJet FX (FJFX) minutiae extractor provides a count of detected minutiae in the finger image. The minutiae count has a bearing on the mated comparison score. The feature computes global quality.

5.2.9.2 MINCNT algorithm

 $Q_{\rm MIN}^{cnt}$ is the number of detected minutiae in the finger image as determined by FJFX.

5.2.10 Minutiae count in center of mass region

5.2.10.1 Description

The FingerJet FX (FJFX) minutiae extractor provides locations of detected minutiae in a finger image. The feature is the minutiae count in a 200×200 pixels local region centered on the center of mass of the detected minutia. The feature computes local quality at the minutiae locations.

5.2.10.2 MINCOM algorithm

 $Q_{\rm MIN}^{com}$ is the number of minutiae occurring within a 200 × 200 pixels local region centered at the center of mass of the locations of all detected minutiae in the finger image as determined by FJFX.

5.2.11 Minutiae quality based on local image mean

5.2.11.1 Description

The FingerJet FX (FJFX) minutiae extractor provides locations of detected minutiae in a finger image. For each minutia location a local quality based on image statistics is computed. The reported quality value is aggregated as the count of local qualities which occurs in the specified range. The feature computes local quality at the minutiae locations.

5.2.11.2 MINMU algorithm

 $\mathbf{Q}_{\mathrm{MIN}}^{mu}$ is computed by first determining the local quality of each minutiae detected by FJFX as

$$\mathbf{Q}_{MIN}^{local_{mu}} = \frac{\mu(\mathbf{I}) - \mu(\mathbf{V})}{\sigma(\mathbf{I})}$$
(27)

where $\mu(I)$ and $\mu(V)$ is arithmetic mean of respectively the finger image and a 32 × 32 pixels local region centered on the minutia and $\sigma(I)$ is the standard deviation of the finger image.

The minutiae quality feature $\boldsymbol{Q}_{\text{MIN}}^{mu}$ is finally computed as the percentage of $\boldsymbol{Q}_{\text{MIN}}^{local_{mu}}$ which have values between 0 and 0,5 as

$$\boldsymbol{Q}_{MIN}^{mu} = \left| \left\{ (x, y) | 0 \le \boldsymbol{Q}_{MIN}^{local_{mu}} < 0.5 \right\} \right|, \quad \text{for} \quad 0 < i \le \boldsymbol{Q}_{MIN}^{cnt}$$
 (28)

5.2.12 Minutiae quality based on local orientation certainty level

5.2.12.1 Description

The FingerJet FX (FJFX) minutiae extractor provides locations of detected minutiae in a finger image. For each minutia location a local orientation certainty level is computed. The reported quality value is aggregated as the count of local qualities which exceed the specified value. The feature computes local quality at the minutiae locations.

5.2.12.2 MINOCL algorithm

 $oldsymbol{Q}_{ ext{MIN}}^{ocl}$ is computed by first determining the local quality of each minutiae detected by FJFX as

$$\mathbf{Q}_{MIN}^{local}{}_{ocl} = \mathbf{Q}_{OCL}^{local}(\mathbf{V}) \tag{29}$$

where $\mathbf{Q}_{\mathrm{OCL}}^{local}(\mathbf{V})$ is the local orientation certainty level (5.2.2) for the 32 × 32 pixels local region \mathbf{V} centered on the minutia.

The minutiae quality feature $m{Q}_{MIN}^{ocl}$ is finally computed as the percentage of $m{Q}_{MIN}^{local_{ocl}}$ which have values greater than 0,8 as

$$\mathbf{Q}_{MIN}^{ocl} = \left| \left\{ (x, y) | \mathbf{Q}_{MIN}^{local_{ocl}} > 0.8 \right\} \right| \quad \text{for} \quad 0 < i \le \mathbf{Q}_{MIN}^{cnt}$$
(30)

5.2.13 Region of interest image mean

5.2.13.1 Description

The region of interest for the finger image is the foreground region of the image containing the fingerprint. The mean image intensity in this area is computed over the set of 32×32 pixels local regions which have a least one pixel contained in the region of interest. The feature computes global quality.

NOTE The quality score is highly correlated with Q_{MU} (5.2.7) and Q_{MMB} (5.2.8).

5.2.13.2 AREA algorithm

- a) Determine the region of interest R (5.2.13.3).
- b) For each 32×32 local region V in I
 - 1) if *V* has at least 1 pixel contained in foreground of *R*, mark the local region as foreground.
- c) Compute $Q_{\mathrm{AREA}}^{\pmb{\mu}}$ as the arithmetic mean of the set of \pmb{V} which are marked as foreground.

5.2.13.3 Determine the Region of Interest

a) Erode the finger image *I* with 5×5 structuring element to obtain *I'*.

ISO/IEC 29794-4:2017(E)

- b) Apply normalized Gaussian blur filter (each weight is divided by the sum of all weights) with kernel size 41×41 and standard deviation of 6,5 to I' to obtain G.
- c) Binarize G using Otsu's method[6] to obtain B.
- d) Apply normalized Gaussian blur filter (each weight is divided by the sum of all weights) with kernel size 91×91 and standard deviation of 14.0 to $\textbf{\textit{B}}$ to obtain $\textbf{\textit{G}}'$.
- e) Binarize **G'** using Otsu's method to obtain **B'**.
- f) Determine the contours of B' using Suzuki's method[\mathbb{Z}] to obtain C.
- g) Regions in \boldsymbol{c} which are surrounded by 0 valued pixels shall be set to 0 valued pixels.
- h) 0 valued pixel regions in C which reach the image border but are not the largest area shall be set to 1 valued pixels.
- i) The resulting binary mask *R* contains the region of interest as a region of 0 valued pixels.

Figure 7 visualizes the processing steps.

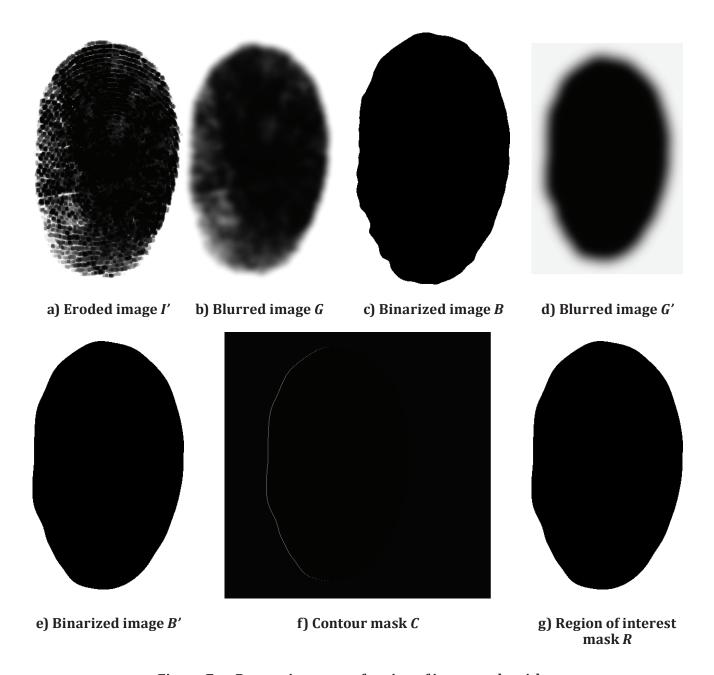


Figure 7 — Processing steps of region of interest algorithm

5.2.14 Region of interest orientation map coherence sum

5.2.14.1 Description

The orientation map coherence sum quantifies the coherence of the estimated finger image orientation field. The coherence map is computed according to coherence method specified in Reference [8]. The feature computes local quality and operates in a block-wise manner.

5.2.14.2 COHSUM algorithm

- a) Compute the pixel-intensity gradient field g of I (5.2.14.3).
- b) Compute the square gradient field as $\boldsymbol{g}_s = \left(g_x^2 g_y^2, 2g_xg_y\right)^T$.
- c) Determine the region of interest R (5.2.13.3).

ISO/IEC 29794-4:2017(E)

- d) For each 16×16 local region V in I
 - 1) if V has at least 1 pixel contained in foreground of R, compute the coherence of V as coh(V)(0), otherwise set coh(V) = 0.
- e) Compute the quality score $oldsymbol{Q}_{ ext{COH}}^{ ext{sum}}$ as the sum of the $coh(oldsymbol{V})$ of all $oldsymbol{V}$.

5.2.14.3 Computing the gradient field

The gradient field $\mathbf{g} = (g_x, g_y)^T$ of pixel intensity I(i,j) of \mathbf{I} is

$$g_{X}(i,j) = I(i+1,j) - I(i-1,j) / 2 \quad \text{for} \quad 1 \le i \le \mathbf{I}_{W} - 1, 0 \le j \le \mathbf{I}_{h}$$

$$g_{X}(0,j) = I(1,j) - I(0,j) \quad \text{for} \quad 0 \le j \le \mathbf{I}_{h}$$

$$g_{X}(\mathbf{I}_{W},j) = I(\mathbf{I}_{W},j) - I(\mathbf{I}_{W}-1,j) \quad \text{for} \quad 0 \le j \le \mathbf{I}_{h}$$
(31)

and

$$g_{y}(i,j) = (I(i,j+1) - I(i,j-1))/2 \quad \text{for} \quad 0 \le i \le \mathbf{I}_{w}, 1 \le j \le \mathbf{I}_{h} - 1$$

$$g_{y}(i,0) = I(i,1) - I(i,0) \quad \text{for} \quad 0 \le i \le \mathbf{I}_{w}$$

$$g_{y}(i,\mathbf{I}_{h}) = I(i,\mathbf{I}_{h}) - I(i,\mathbf{I}_{h} - 1) \quad \text{for} \quad 0 \le i \le \mathbf{I}_{w}$$
(32)

where I_w and I_h are respectively the width and height of I in pixels.

5.2.14.4 Computing the coherence of a local region

The coherence of a local region V is computed from its pixel-intensity gradient field g_s as

$$coh(\mathbf{V}) = \frac{\left|\sum g_s(i,j)\right|}{\sum \left|g_s(i,j)\right|} \tag{33}$$

where $| \cdot |$ denotes the Euclidean norm and the sums are taken over all pixels in V.

5.2.15 Region of interest relative orientation map coherence sum

5.2.15.1 Description

The relative orientation map coherence sum is the average of local region orientation coherence as determined by <u>5.2.14</u>. The feature computes local quality and operates in a block-wise manner.

5.2.15.2 COHREL algorithm

- a) Compute Q_{COH}^{sum} and store the number of local regions V which have at least one pixel contained in R as n (5.2.14).
- b) Compute $Q_{\text{COH}}^{rel} = \frac{Q_{\text{COH}}^{sum}}{n}$.

5.2.16 Quality feature vector composition

5.2.16.1 Description

Quality features specified in <u>5.2.2</u> to <u>5.2.6</u> provide a map of values for local regions in the finger image. The quality features specified in <u>5.2.7</u> to <u>5.2.15</u> provide scalar quality values for the finger image.

The specified features shall be composed such that a fixed length feature vector is obtained for use by a classification system. Thus, the features in 5.2.2 to 5.2.6 are aggregated using arithmetic mean

specified in <u>5.2.16.2</u>, standard deviation specified in <u>5.2.16.3</u> and histogram specified in <u>5.2.16.4</u> for inclusion in the final feature vector specified in <u>5.2.16.5</u>.

5.2.16.2 Mean of local quality values

The mean quality value $m{Q}_{
m AREA}^{\,\mu}$ over an N imes M matrix of local quality values $m{Q}_{
m QNAME}^{\,local}$ is computed as

$$\boldsymbol{Q}_{QNAME}^{\mu} = \frac{1}{N*M} \sum_{i=1}^{N} \sum_{j=1}^{M} \boldsymbol{Q}_{QNAME}^{local}$$
(34)

where *qname* is one of OCL, LCS, FDA, RVU, OFL corresponding to the local quality values computed for orientation certainty (5.2.2), local clarity score (5.2.3), frequency domain analysis score (5.2.4), ridge valley uniformity (5.2.5), orientation flow (5.2.6).

This yields the arithmetic mean aggregated quality scores $\mathit{Q}_{\mathrm{OCL}}^{\,\mu}$, $\mathit{Q}_{\mathrm{LCS}}^{\,\mu}$, $\mathit{Q}_{\mathrm{FDA}}^{\,\mu}$, $\mathit{Q}_{\mathrm{OFL}}^{\,\mu}$, $\mathit{Q}_{\mathrm{OFL}}^{\,\mu}$

5.2.16.3 Standard deviation of local quality values

The standard deviation Q_{ONAME}^{μ} over an $N \times M$ matrix of local quality values $Q_{\mathrm{ONAME}}^{local}$ is computed as

$$Q_{QNAME}^{\sigma} = \left(\frac{1}{N*M-1} \sum_{i=1}^{N} \sum_{j=1}^{M} \left(\mathbf{Q}_{QNAME}^{local}\left(i,j\right) - Q_{QNAME}^{\mu} \right)^{2} \right)^{\frac{1}{2}}$$
(35)

where *qname* is one of OCL, LCS, FDA, RVU, OFL corresponding to the local quality values computed for orientation certainty (5.2.2), local clarity score (5.2.3), frequency domain analysis score (5.2.4), ridge valley uniformity (5.2.5), orientation flow (5.2.6).

This yields the standard deviation aggregated quality scores $\mathit{Q}_{\mathrm{OCL}}^{\mu}$, $\mathit{Q}_{\mathrm{LCS}}^{\mu}$, $\mathit{Q}_{\mathrm{RVU}}^{\mu}$, $\mathit{Q}_{\mathrm{OFL}}^{\mu}$

5.2.16.4 Histogram of local quality

Local quality values from orientation certainty (5.2.2), local clarity score (5.2.3), frequency domain analysis score (5.2.4), ridge valley uniformity (5.2.5), orientation flow (5.2.6) shall be represented as fixed-length histograms with 10 bins to capture the distribution of local qualities.

The boundaries defining each bin for each of the features are specified as

 $B_{FDA} = \{-\infty; 0,268\ 00; 0,304\ 00; 0,330\ 00; 0,355\ 00;$

 $0,380\ 00;0,407\ 00;0,440\ 00;0,500\ 00;1,000\ 00;\infty$

 $B_{LCS} = \{-\infty; 0,000\ 00; 0,700\ 00; 0,740\ 00; 0,770\ 00;$

 $0,790\ 00;0,810\ 00;0,830\ 00;0,850\ 00;0,870\ 00;\infty$

$$B_{OCL} = \{-\infty; 0.337\ 00; 0.479\ 00; 0.579\ 00; 0.655\ 00;$$
(36)

 $0,716\ 00;0,766\ 00;0,810\ 00;0,852\ 00;0,898\ 00;\infty$

 $0,115\ 00;0,171\ 80;0,256\ 90;0,475\ 80;0,748\ 00;\infty$

 $B_{RVU} = \{-\infty, 0,500\ 00, 0,667\ 00, 0,800\ 00, 1,000\ 00\}$

 $1,250\ 00;1,500\ 00;2,000\ 00;24,000\ 0;30,000\ 0;\infty$

For each of FDA, LCS, OCL, OFL, RVU, a histogram is computed using the specified bin boundaries where the *i*th bin in the histogram is given by the interval

$$\begin{pmatrix} B_{\mathbf{Q}}^{i}, B_{\mathbf{Q}}^{i+1} \end{pmatrix} , for \quad 1 = i$$

$$\begin{bmatrix} B_{\mathbf{Q}}^{i}, B_{\mathbf{Q}}^{i+1} \end{pmatrix} , for \quad 1 < i \le |B_{\mathbf{Q}}| \tag{37}$$

The *i*th interval includes the value of $B_{\mathbf{Q}}^{i}$ on the left and excludes the value of $B_{\mathbf{Q}}^{i+1}$ on the right when $1 < i \le |B_{\mathbf{Q}}|$. In this document, $|B_{\mathbf{Q}}| = 10$.

The histograms of local qualities are specified according to their bin boundaries as defined in Formulae (36) and (37) where the *i*th bin in the histogram contains the cardinality of the multiset that contains values bounded by the histogram boundaries

$$Q_{\text{FDA}}^{i} = \left| \left\{ (x, y) | B_{\text{FDA}}^{i} \leq \mathbf{Q}_{\text{FDA}}^{local} < B_{\text{FDA}}^{i+1} \right\} \right| , for 1 \leq i \leq \left| B_{\text{FDA}} \right|,$$

$$Q_{\text{LCS}}^{i} = \left| \left\{ (x, y) | B_{\text{LCS}}^{i} \leq \mathbf{Q}_{\text{LCS}}^{local} < B_{\text{LCS}}^{i+1} \right\} \right| , for 1 \leq i \leq \left| B_{\text{LCS}} \right|,$$

$$Q_{\text{OCL}}^{i} = \left| \left\{ (x, y) | B_{\text{OCL}}^{i} \leq \mathbf{Q}_{\text{OCL}}^{local} < B_{\text{OCL}}^{i+1} \right\} \right| , for 1 \leq i \leq \left| B_{\text{OCL}} \right|,$$

$$Q_{\text{OFL}}^{i} = \left| \left\{ (x, y) | B_{\text{OFL}}^{i} \leq \mathbf{Q}_{\text{OFL}}^{local} < B_{\text{OFL}}^{i+1} \right\} \right| , for 1 \leq i \leq \left| B_{\text{RVU}} \right|,$$

$$Q_{\text{RVU}}^{i} = \left| \left\{ (x, y) | B_{\text{RVU}}^{i} \leq \mathbf{Q}_{\text{RVU}}^{local} < B_{\text{RVU}}^{i+1} \right\} \right| , for 1 \leq i \leq \left| B_{\text{RVU}} \right|.$$

$$(38)$$

The histogram for a single feature represented by its bins is written as

$$\mathbf{Q}_{QNAME} = \left\{ Q_{QNAME}^{i} \right\}, \text{for } 1 \le i \le B_{QNAME}$$
(39)

where *qname* is one of OCL, LCS, FDA, RVU, OFL.

This yields the histogram feature vectors $m{Q}_{
m OCL}$, $m{Q}_{
m LCS}$, $m{Q}_{
m FDA}$, $m{Q}_{
m RVU}$, $m{Q}_{
m OFL}$

5.2.16.5 ISO 29794-4 quality feature vector

This document's quality feature vector is specified as

$$\begin{aligned} \boldsymbol{Q}_{29794-4} &= \{ & Q_{\text{OCL}}^{\mu}, Q_{\text{LCS}}^{\mu}, Q_{\text{FDA}}^{\mu}, Q_{\text{RVU}}^{\mu}, Q_{\text{OFL}}^{\mu}, \\ & Q_{\text{OCL}}^{\sigma}, Q_{\text{CCS}}^{\sigma}, Q_{\text{FDA}}^{\sigma}, Q_{\text{RVU}}^{\sigma}, Q_{\text{OFL}}^{\sigma}, \\ & \boldsymbol{Q}_{\text{OCL}}, \boldsymbol{Q}_{\text{LCS}}, \boldsymbol{Q}_{\text{FDA}}, \boldsymbol{Q}_{\text{RVU}}, \boldsymbol{Q}_{\text{OFL}}, \\ & \boldsymbol{Q}_{\text{OU}}, \boldsymbol{Q}_{\text{LCS}}, \boldsymbol{Q}_{\text{FDA}}, \boldsymbol{Q}_{\text{RVU}}, \boldsymbol{Q}_{\text{OFL}}, \\ & Q_{\text{MU}}, Q_{\text{MMB}}, Q_{\text{COH}}^{rel}, Q_{\text{COH}}^{sum}, Q_{\text{AREA}}^{\mu}, \\ & Q_{\text{MIN}}^{cnt}, Q_{\text{MIN}}^{com}, Q_{\text{MIN}}^{mu}, Q_{\text{MIN}}^{ocl}, \} \end{aligned}$$

$$(40)$$

5.3 Non-normative quality metrics

5.3.1 General

<u>5.3</u> specifies non-normative finger image quality assessment algorithms.

5.3.2 Radial power spectrum

5.3.2.1 Description

The radial power spectrum is a measure of maximal signal power in a defined frequency band of the global radial Fourier spectrum. Ridges can be locally approximated by means of a single sine wave, hence high energy concentration in a narrow frequency band corresponds to consistent ridge structures.

Since the ridges of a finger image can be locally approximated by one sine wave, large value of sine wave energy can represent the strong ridges. The robustness of the ridge structure can be used to measure the finger image quality. F is decided as the maximum Radial Fourier spectrum value within the reasonable Fourier domain. The reasonable Fourier domain refers to the region of neither the highest nor the lowest frequency. The higher the value of F, the better is the finger image quality.

5.3.2.2 Variables

Name	Default	Description
$r_{ m min}$	0,143 cycles/pixel	Lower bound of frequency band
$r_{\rm max}$	0,077 cycles/pixel	Upper bound of frequency band
Δ_r		Sampling step between annular bands in the frequency spectrum
θ	180	Degrees of the spectrum to consider

5.3.2.3 Algorithm

- a) Compute the magnitude of the 2D-DFT F(u,v) of input image.
- b) Transform F(u,v) into polar coordinates and normalize to the range of [0, 1].
- c) Determine the maximum energy to compute Q_{POW} (5.3.2.5).

5.3.2.4 Magnitude of frequency bands polar coordinates

The magnitude of the annular band between r and $r + \Delta_r$ in the polar Fourier spectrum $F(\alpha,r)$ is computed as shown in Formula (41):

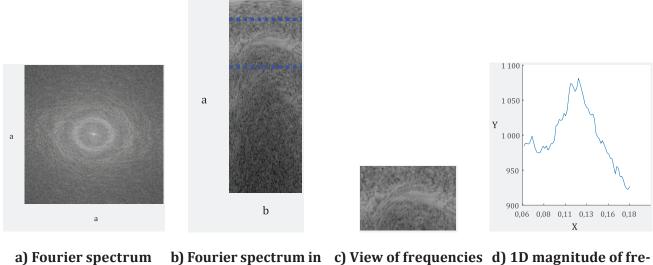
$$J(r) = \frac{\sum_{\alpha=0}^{\pi} \sum_{r}^{r+\Delta_{r}} F(\alpha, r)}{\sum_{\alpha=0}^{\pi} \sum_{r_{\min}}^{r_{\max}} F(\alpha, r)}$$
(41)

where

 α is the angle;

r is the radius.

 $F(\alpha,r)$ is the Spectrum f(p,q) representation in polar coordinate system (α,r) , see Figure 8.



a) Fourier spectrum of input

- polar coordinate system
- c) View of frequer of interest
- d) 1D magnitude of frequencies of interest

Key

- X cycles/pixel
- Y magnitude
- a Cycles/pixel.
- b Radians.

Figure 8 — Processing steps of radial power spectrum algorithm

5.3.2.5 Determine quality score from energy distribution

The quality feature Q_{POW} is found as

$$Q_{POW} = \max_{r \in [\Delta r]} |J(r)| \tag{42}$$

NOTE The Radial Power Spectrum quality feature is spatial sampling rate dependent. The given defaults assume 196,85 pixel per centimetre (500 pixel per inch).

5.3.3 Gabor quality score

5.3.3.1 Feature description

The Gabor quality feature operates on a per-pixel basis by calculating the standard deviation of the Gabor filter bank responses^[9]. The size of the filter bank is used to determine a number of filters oriented evenly across the half circle. The strength of the response at a given location corresponds to the agreement between filter orientation and frequency in the location neighbourhood. For areas in the fingerprint image with a regular ridge-valley pattern there will be a high response from one or a few filter orientations. In areas containing background or unclear ridge-valley structure the Gabor response of all orientations will be low and constant.

5.3.3.2 Variables

Name	Default	Description
σ_{X}	6	2D Gaussian standard deviation in x-direction
σ_y	6	2D Gaussian standard deviation in y-direction
n	4	Size of filter bank (orientations of the Gabor wave)
f	0,1	Gabor filter frequency
θ	_	An orientation of a Gabor filter

5.3.3.3 Algorithm

- a) Convolve input image with a 2D Gaussian kernel with σ =1 and subtract it from the input image I to give \hat{I} .
- b) Compute the Gabor response of \hat{I} for each orientation θ .
- c) Convolve the magnitude (complex modulus) of each Gabor response with a 2D Gaussian kernel with σ = 4.
- d) Compute the standard deviation of the Gabor magnitude response values at each location yielding a map of standard deviations.
- e) Sum the map of standard deviations and normalize according to number of sample points to produce the final Gabor quality score.

Figure 9 visualizes the processing steps.

5.3.3.4 Gabor filter

The general form of the complex 2D Gabor^[10] filter h_{Cx} in the spatial domain is given by Formula (43):

$$h_{Cx}\left(x,y;f,\theta,\sigma_{x},\sigma_{y}\right) = \exp\left(-\frac{1}{2}\left(\frac{x_{\theta}^{2}}{\sigma_{x}^{2}} + \frac{y_{\theta}^{2}}{\sigma_{y}^{2}}\right) \exp\left(j2\pi f x_{\theta}\right)\right)$$
(43)

where

$$x_{\theta} = x\sin\theta + y\cos\theta \tag{44}$$

$$y_{\theta} = x\cos\theta - y\sin\theta \tag{45}$$

and f is the frequency (cycles/pixel) of the sinusoidal plane wave along the orientation θ . The size of the Gaussian smoothing window is determined by σ_{x} , σ_{v} .

The filter bank size n is used to compute the differently oriented Gabor filters composing the filter bank. Computing θ given n is done as:

$$\theta = \frac{k-1}{n\pi}, k = 1, \dots, n \tag{46}$$

NOTE The Gabor quality feature is spatial sampling rate dependent. The given defaults assume 196,85 pixel per centimetre (500 pixel per inch).

5.3.3.5 Computing the Gabor quality score from the Gabor filter response

Let G be a matrix with standard deviations of local responses resulting from convolution of I and Gabor filter of orientation n. The Gabor quality Q_{GAB} is computed as:

$$Q_{GAB} = \frac{1}{X * Y} \sum_{x} \sum_{y} G(x, y) \tag{47}$$

responses at each pixel

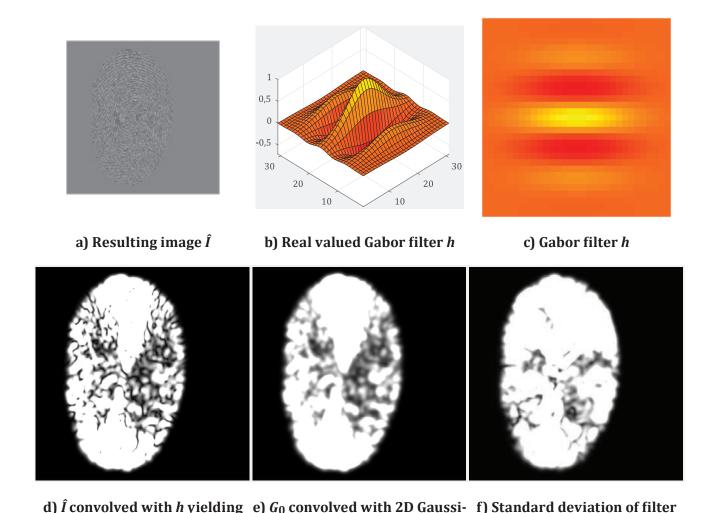


Figure 9 — Processing steps of Gabor algorithm

an kernel

5.4 Unified quality score

matrix of filter responses G_0

5.4.1 Methodology for combining quality metrics

In order to obtain a single or unified output from several or all the quality metrics described in the earlier clauses, it is necessary to combine the values of the quality metrics described above and produce a single scalar quality score as required in the quality field. Each of the quality metrics shall be normalized to the range between 0 and 100 prior to combining them. Combining quality metrics shall be done such that the overall quality score is predictive of performance. There are various methods that can be used to combine all the quality metrics, e.g. weighted averaging, the use of pattern classifiers and other nonlinear computations.

5.4.2 Training method

Pattern classifiers are mathematical models that can intelligently learn a concept and predict an output when presented with new and even unseen samples. To apply pattern classification to combine the finger image quality analysis metrics, it is necessary to train the pattern classifier by providing finger images with the values for all the quality metrics computed and the overall quality scores for each sample. Once the pattern classifier is well-trained, given the values of the quality metrics, it will be able to provide an overall quality score for the finger image.

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The feature vector, $Q_{29797-4}$ (5.2.16.5), will be the input to the pattern classifier. Training the pattern classifier could be performed using a corpus of finger images with pre-assigned quality categories or scores such as the QSND corpus, on the output of one or many quality algorithms. A detailed approach to establish the QSND and the minimum number of samples required can be found in ISO/IEC 29794-1. For all the samples in the corpus, the feature vectors are computed. They are then paired with the quality category or score and fed into the pattern classifier for training.

With the feature vector specified in <u>5.2.16.5</u>, a Random Forest shall be trained for binary classification, where Class 0 represents images of very low utility and Class 1 represents images of very high utility. The trained random forest outputs class membership along with its probability score. This score is the probability that a given image belongs to class 1 multiplied by 100 and rounded to its closest integer.

The number of images chosen for the training process shall conform to ISO/IEC 29794-1. The training set shall be chosen such that:

- a) Class 1 (or high utility) consists of images with NFIQ $1.0^{[11]}$ value of 1 (with activation score >0,7) and genuine score in the 90^{th} percentile for each of the comparison score providers.
- b) Class 0 (or low utility) consists of images with NFIQ 1.0 value of 5 (with activation score >0,9) and genuine score smaller than a threshold value that corresponds to false match rate of 1 in 10 000, i.e. false reject at false match rate of 0,000 1.

6 Finger image quality data record

6.1 Binary encoding

In binary data records, quality data shall be encoded as described in <u>Table 1</u>.

Byte #	Name	Length	Valid values	Description + Notes
0	Number of Quality	1 byte	0 to 255	This field is followed by the number of 5-byte Quality Blocks reflected by its value.
	Blocks			A value of zero (0) means that no attempt was made to assign a quality score. In this case, no Quality Blocks are present.
1	Quality score	1 byte	0 to 100, 255	Quality score of the metric identified by the Quality Algorithm Identifier (QAID) in bytes 4 and 5 of this Quality Block.
				If quality score is equal to 255 (FF $_{\rm Hex}$), an attempt to calculate a quality score has failed.
2-3	Quality Algorithm Vendor Identifier	lgorithm endor	0 to 65535 257 (0101 _{HEX}) for standard quality.	This field shall contain the identifier of the vendor whose algorithm was used to compute quality. Quality algorithm vendor identifier shall be registered with IBIA or other approved registration authority as a CBEFF biometric organization in accordance with CBEFF vendor ID registry procedures in ISO/IEC 19785-2. A value of all zeros shall indicate that the value for this field is unreported.
				SC37 vendor ID (257 or 0101HEX) shall be used if and only if an SC 37 approved reference implementation is used to compute the quality score.
				The reference implementation is posted at https://github.com/usnistgov/NFIQ2 with the tag NFIQ2.0_29794-4_edition_2017.
4-5	Quality Algorithm Identifier (QAID)	2 bytes	1 to 65535	The quality algorithm identifier shall be encoded in two bytes. A value of all zeros is not permitted.
				If encoding standard quality metrics defined in this document, the quality algorithm identifiers defined in Table 2 shall be used.
				For encoding of quality components not defined in the specific modality parts the quality algorithm identifier shall be assigned by the vendor or an approved registration authority.

Table 1 — Finger image quality data record structure

Byte 1/2-3/4-5: 5-byte Quality Block [0 or more]

Quality scores should always be placed within the quality record of the biometric data interchange record (BDIR) as defined in ISO/IEC 19794-x associated with the sample. CBEFF quality fields should not be used in place of 19794 quality fields but rather as supplementary data. The prescribed use of CBEFF quality fields may be supplied by each CBEFF patron format standard and is beyond the scope of this document. Multiple quality scores calculated by the same algorithm (same quality algorithm vendor identifier and same quality algorithm identifier) shall not be present in a single BDIR.

6.2 XML encoding

In XML documents, quality data shall be encoded as described in the following XML type definitions.

```
<xs:element name="Quality" type="QualityType" maxOccurs="255"/>
  </xs:sequence>
</xs:complexType>
<xs:complexType name="QualityType">
  <xs:sequence>
    <xs:element name="Algorithm" type="RegistryIDType"/>
    <xs:choice>
      <xs:element name="Score" type="QualityScoreType"/>
      <xs:element name="QualityCalculationFailed">
        <xs:complexType/>
      </xs:element>
    </xs:choice>
  </xs:sequence>
</xs:complexType>
<xs:simpleType name="QualityScoreType">
  <xs:restriction base="xs:unsignedByte">
    <xs:minInclusive value="0"/>
    <xs:maxInclusive value="100"/>
  </xs:restriction>
</xs:simpleType>
```

6.3 Quality algorithm identifiers

07

08

09

10

The owner of the quality algorithms defined in this document is ISO/IEC JTC 1/SC 37. Its organization identifier is 257 (101_{Hex}). Table 2 lists the quality algorithm identifiers for the quality metrics defined in this document.

If a unified quality score is calculated and reported in a finger image quality data record, the normative quality metrics defined in <u>5.2</u> have to be calculated first. The values of the normative quality metrics defined in <u>5.2</u> may but need not be reported in the finger image quality data record. Calculation and reporting of the non-normative quality metrics defined in <u>5.3</u> is optional.

NOTE The unified quality score consists of the features in <u>Formula (41)</u> in <u>5.2.16.5</u>.

Quality Quality Algorithm Governing algorithm algorithm Quality metric feature subclause + identifier identifier name description decimal Hex 01_{Hex} 01 Unified quality score ($Q_{29794-4}$) Finger image quality score (5.4) 02_{Hex} 02 OCL 5.2.2, 5.2.16.2 Mean of local orientation certainty level (Q_{OCL}^{μ}) Standard deviation of local orientation certain- 03_{Hex} 03 OCL 5.2.2, <u>5.2.16.3</u> ty level $(Q_{\text{OCL}}^{\sigma})$ LCS 04 <u>5.2.3</u>, <u>5.2.16.2</u> 04_{Hex} Mean of local clarity score (Q_{LCS}^{μ}) 05 LCS <u>5.2.3</u>, <u>5.2.16.3</u> 05_{Hex} Standard deviation of local clarity score (Q_{LCS}^{σ}) 06 **FDA** 06_{Hex} <u>5.2.4</u>, <u>5.2.16.2</u> Mean of local frequency domain analysis ($Q_{ extsf{FDA}}^{\mu}$)

Standard deviation of local frequency domain

Mean of local ridge valley uniformity (Q_{RVU}^{μ}) Standard deviation of local ridge valley

Mean of local orientation flow (Q_{OFI}^{μ})

analysis ($Q_{ extsf{FDA}}^{\sigma}$)

uniformity $(Q_{\text{RVII}}^{\sigma})$

Table 2 — Quality metric identifier

 07_{Hex}

 08_{Hex}

 09_{Hex}

 $0A_{Hex}$

<u>5.2.4</u>, <u>5.2.16.3</u>

5.2.5, 5.2.16.2

5.2.5, 5.2.16.3

5.2.6, 5.2.16.2

FDA

RVU

RVU

OFL

 Table 2 (continued)

Quality algorithm identifier Hex	Quality algorithm identifier decimal	Quality metric	Algorithm feature name	Governing subclause + description
0B _{Hex}	11	Standard deviation of orientation flow $(Q_{ m OFL}^{\sigma})$	OFL	5.2.6, 5.2.16.3
0C _{Hex}	12	$MU(Q_{MU})$	MU	5.2.7
0D _{Hex}	13	MMB (Q _{MMB})	MMB	5.2.8
0E _{Hex}	14	Minutiae count (Q_{MIN}^{cnt})	MINCNT	5.2.9
0F _{Hex}	15	Minutiae count in center of mass $(Q_{ ext{MIN}}^{com})$	MINCOM	5.2.10
10 _{Hex}	16	Minutiae quality based on image mean $(Q_{ ext{MIN}}^{ ext{mu}})$	MINMU	5.2.11
11 _{Hex}	17	Minutiae quality based on orientation certainty level ($Q_{ m MIN}^{ocl}$)	MINOCL	5.2.12
12 _{Hex}	18	Region of interest image mean $(Q_{ ext{AREA}}^{\mu})$	AREA	5.2.13
13 _{Hex}	19	Region of interest orientation map coherence sum (Q_{COH}^{sum})	COHSUM	5.2.14
14 _{Hex}	20	Region of interest relative orientation map coherence sum $(Q_{ m COH}^{rel})$	COHREL	5.2.15
15 _{Hex}	21	Radial power spectrum (QPOW)	POW	5.3.2
16 _{Hex}	22	Gabor quality score (Q _{GAB})	GAB	5.3.3

Annex A

(normative)

Conformance test assertions

A.1 Overview

This document specifies terms and quantitative methodologies relevant to characterizing the quality of finger images and to assessing their potential for high confidence biometric match decisions.

The objective of this document cannot be completely achieved until biometric products can be tested to determine whether they conform to those specifications. Conforming implementations are a necessary prerequisite for achieving interoperability among implementations; therefore there is a need for a standardised conformance testing methodology, test assertions, and test procedures as applicable to specific modalities addressed by this document. The test assertions will cover as much as practical of this document's requirements (covering the most critical features), so that the conformity results produced by the test suites will reflect the real degree of conformity of the implementations to ISO/IEC 29794-4 finger image quality data records. This is the motivation for the development of this conformance testing methodology.

This annex is intended to specify elements of conformance testing methodology, test assertions, and test procedures as applicable to this document.

A.2 Conformance test set

To verify conformance, implementations of quality assessment algorithms claiming conformance with this document should run on all images in the conformance test set. In order to conform, no output value shall differ from that of the reference implementation by more than 1 %.

The finger images in the conformance test set are selected from databases DB1 and DB3 of FVC2000[12] and database DB1 of FVC2002[13].

NFIQ 2.0, the source code of which is available from https://github.com/usnistgov/NFIQ2 is the reference implementation for this document.

Table A.1 — Quality metric values for conformance test set

$Q_{\rm COH}^{rel}$	0,627	0,567	0,614	0,635	0,596	0,603	0,673	0,539	0,564	989'0 86	0,731	0,723	0,574	0,793	0,651	0,508	0,633	0,752	0,778
Q sum		177,01	179,39	230,78	399,33	425,99	212,86	317,28	290,62	195,07	264,94	225,59	220,45	261,83	204,03	364,03	216,13	275,5	212,57
$(Q^{\mu}_{ m AREA})$	141,88	189,8	193,68	187,89	150,28	154,88	196,39	158,29	163,87	196,26	152,12	145,32	184,12	139,95	199,99	143,92	204,37	147,05	167,09
Qocl	0,29231	0,26	0,369 86	0,3617	0,203 7	0,213 59	0,58824	0,096	0,13636	0,5	0,51613	0,613 64	0,162 79	0,674 42	0,487 18	0,053	0,422 22	0,595 74	0,73469
Q _{min}	0,769 23	0,38	0,479 45	0,212 77	0,61111	6 699'0	0,55882	8 869'0	0,727 27	0,4	0,35484	0,25	0,34884	0,39535	0,615 38	0,60638	0,533 33	0,340 43	0,34694
Q com	25	30	43	33	35	38	30	28	36	25	38	30	30	24	33	29	25	22	37
Quin	65	20	73	47	108	103	34	83	99	40	62	44	43	43	39	94	45	47	49
Оммв	162,71	214,45	218,21	209,58	171,1	170,93	202,69	184,83	197,69	221,24	186,89	186,56	204,65	183,31	206,81	155,97	221,93	181,45	182,44
Q _{MU}	162,85	213,76	216,56	206,52	171,2	171,0	200,44	184,81	197,72	219,83	183,61	179,95	202,51	180,28	204,55	156,11	219,97	177,99	179,08
Q OFI.	0,330	0,249	0,370	0,257	0,301	0,256	0,246	0,326	0,314	0,252	0,254	0,346	0,375	0,267	0,268	0,295	0,225	0,231	0,281
(Q_{OFL}^{μ})	0,2061	0,18448	0,27658	0,165 16	0,20876	0,17614	0,17773	0,22035	0,22853	0,15683	0,18264	0,262 18	0,257 66	0,18192	0,20034	0,214 18	0,1528	0,12999	0,2035
Q GRVII	0,676	0,644	0,645	0,381	0,765	0,606	0,483	0,839	0,919	979,0	0,707	0,631 1	0,3688	0,456	1,008 4	968'0	0,602	0,952	0,575
(Q^{μ}_{RVU})	1,1465	1,087 9	1,0585	1,0486	1,143 4	1,115 4	1,1249	1,0998	1,118 7	1,142 2	1,1867	1,1461	1,0349	1,0409	1,2053	1,1393	1,09	1,2051	1,0853
$Q_{\mathrm{FDA}}^{\sigma}$	0,175	0,1462	0,134	0,128	0,202	0,190	0,159	0,227	0,155	0,122	0,138	0,168	0,114	0,1464	0,197	0,213	0,120	0,162	0,207
(Q_{FDA}^{μ})	0,4641	0,420	0,397	0,432	0,463	0,462	0,418	0,496	0,458	0,407	0,409	0,423	0,438	0,421	0,418 4	0,473	0,418	0,410	0,451
$Q_{\rm LCS}^{\sigma}$	6 4	0,213	0,279 7	0,184	0,198	0,213	0,184	0,216	0,157	0,311	0,228	0,247	0,105	0,110	0,246	0,233	0,209	0,224	0,207
$(Q_{\rm LCS}^{\mu})$	0,7792	0,749	0,700	0,775	0,766	0,763	0,776	0,767	0,793	0,677 2	0,760	0,758	0,812	0,831	0,719	0,7263	0,755	0,773 5	0,778
Q _{OCI.}	0,1786	0,152 1	0,18692	0,12564	0,18121	0,159 09	0,13167	0,227 93	0,247 58	0,16481	0,11331	0,18633	0,16772	0,17987	0,158 07	0,16431	0,157 95	0,125 26	0,13113
$(Q_{\rm OCL}^{\mu})$	0,71789	0,699 26	0,720 02	0,7591	0,700 7	0,714 75	0,7902	0,647 79	0,64196	0,757	0,837.87	0,803 25	0,698 28	0,809 43	0,755 95	0,63043	0,735 71	0,841 79	0,83135
Q ₂₉₇₉₄₋₄		68	88	88	87	87	98	98	82	82	84	84	83	83	82	82	81	81	08
Image identifier	FVC2000/ Db3/25_3	FVC2002/ Db1/49_6	FVC2002/ Db1/51_4	FVC2002/ Db1/89_1	FVC2000/ Db3/108_3	FVC2000/ Db3/108_4	FVC2000/ Db1/69_6	FVC2000/ Db3/26_2	FVC2000/ Db3/24_4	FVC2002/ Db1/29_2	FVC2002/ Db1/51_5	FVC2002/ Db1/55_6	FVC2002/ Db1/110_7	FVC2002/ Db1/14_6	FVC2000/ Db1/69_8	FVC2000/ Db3/108_6	FVC2002/ Db1/105_2	FVC2002/ Db1/106_5	FVC2000/ Db1/36_6

Table A.1 (continued)

	72	68 42	.14 96	75 33	_	08	97 23	40 21	96 22	62 16	.01 86	110	02 01	2	.41 29	78	96 61	84 92	93
Q_{COH}^{rel}	0,776	0,7	0,714 96	0,7	0,7197	8′0	9'0	0,7	9'0	0,7	0,7	8'0	0,7	0,7652	0,7	0,7	0,7	0,7	0,7
$Q_{\rm COH}^{sum}$	187,91	220,54	179,45	183,75	184,96	194,9	156,18	205,04	165,7	218,74	195,82	230,12	197,27	211,96	186,8	252,17	238,19	183,67	175,27
$(Q_{ m AREA}^{\mu})$	186,48	183,02	186,23	167,7	196,84	170,39	200,91	166,97	220,45	181,09	148,8	173,2	178,67	185,73	175,41	176,86	181,17	175,13	174,57
$Q_{ m MIN}^{ocl}$	0,6	0,512 2	0,382 98	0,6875	0,64	0,772 73	0,29167	0,54167	0,45	0,58824	0,61111	0,682 93	0,625	0,58824	0,70588	0,69231	0,812 5	0,5	0,566 67
Q _{MIN}	0,566 67	0,390 24	0,2766	0,333 33	0,32	0,227 27	0,375	0,16667	0,35	0,470 59	0,361 11	0,609 76	0,28125	0,196 08	0,323 53	0,307 69	0,625	0,333 33	0,366 67
Q com	22	29	37	35	25	36	24	17	20	25	28	32	28	43	33	17	8	29	25
$Q_{ m MIN}^{cnt}$	30	41	47	48	25	44	24	24	20	34	36	41	32	51	34	26	16	36	30
$Q_{ m MMB}$	197,49	191,78	197,64	186,25	207,93	187,12	211,58	184,67	229,7	192,94	170,75	185,21	193,37	199,61	190,13	184,1	193,14	189,45	193,46
Оми	196,62	190,08	195,53	182,41	205,88	183,69	210,21	180,84	228,96	190,83	166,07	183,05	189,4	196,68	188,11	181,75	190,73	187,25	190,06
$Q_{ m OFL}^{\sigma}$	0,217	0,255	0,2632	0,283	0,247	0,2661	0,251	0,273	0,250	0,307	0,403	0,235	0,3178	0,234	0,244	0,239	0,304	0,2189	0,204
$(Q_{ m OFL}^{\mu})$	0,15538	0,180 95	0,1896	0,21273	0,18153	0,185 48	0,213 76	0,190 32	0,22194	0,20666	0,275 98	0,146 59	0,23467	0,188 63	0,1783	0,155 16	0,19413	0,141 76	0,142
$Q_{ m RVU}^{\sigma}$	0,579	0,452	0,757	0,693	1,168 6	1,046 6	1,255	0,809	1,121 6	1,053	0,891	0,645	0,762	0,9816	0,905	0,664	0,747	0,925	0,890
(Q^{μ}_{RVU})	1,089	1,031 5	1,1563	1,102 9	1,2668	1,162 1	1,222 7	1,155 1	1,250 7	1,211 1	1,142 3	1,087 5	1,161 5	1,165 6	1,1541	1,115 2	1,1928	1,195 5	1,147 2
$Q_{\mathrm{FDA}}^{\sigma}$	0,170	0,208	0,182 3	0,242	0,275	0,228	0,261	0,273	0,283	0,240	0,251	0,2458	0,241	0,282	0,249	0,207	0,178	0,224	0,250
(Q_{FDA}^{μ})	0,452	0,473 6	0,422	0,498	0,495	0,474	0,473 9	0,526	0,475	0,487	0,506	0,5033	0,485	0,525	0,484	0,466	0,458	0,467	0,529
$Q_{\mathrm{LCS}}^{\sigma}$	0,143	0,203	0,161	0,316	0,258	0,287	0,283	0,204	0,293	0,187	0,228 5	0,254	0,248	0,261	0,272	0,165	0,141	0,219	0,105
(Q_{LCS}^{μ})	0,8213	0,772	0,7808	0,681	0,710	0,717	0,666	0,765	0,653	0,793	0,750	0,749	0,737	0,722	0,711	0,814	0,8314	0,750	90,808
$Q_{ m OCL}^{\sigma}$	0,125 11	0,147 77	0,12432	0,1399	0,183 44	0,12801	0,183 7	0,1246	0,2692	0,11897	0,175 58	0,109 76	0,16994	0,17056	0,16493	686	0,117 62	0,142 7	0,141 62
$(Q_{0\text{CL}}^{\mu})$	0,81121	0,81404	0,80131	0,81388	0,756 28	0,835 49	0,729 08	0,817 94	0,6627	0,833	0,7753	0,85455	0,768 12	0,802 28	0,773 35	0,85626	0,85178	0,80984	0,815 22
Q ₂₉₇₉₄₋₄	80	79	62	78	78	77	77	92	92	75	75	74	74	73	73	72	72	71	71
Image identifier	FVC2000/ Db1/9_5	FVC2000/ Db1/104_8	FVC2000/ Db1/69_5	FVC2000/ Db1/28_5	FVC2000/ Db1/58_6	FVC2000/ Db1/28_6	FVC2000/ Db1/33_1	FVC2000/ Db1/11_5	FVC2000/ Db1/19_5	FVC2000/ Db1/104_6	FVC2000/ Db1/105_3	FVC2000/ Db1/2_5	FVC2000/ Db1/24_5	FVC2000/ Db1/25_8	FVC2000/ Db1/28_2	FVC2000/ Db1/101_7	FVC2000/ Db1/101_8	FVC2000/ Db1/20_6	FVC2000/ Db1/23_2

Table A.1 (continued)

	00	80	06	33	98	86 36	80	33	68	83	78	35	60 31	50 81	8	67 39	01		77 05
Q_{COH}^{rel}	8′0	0,7	8′0	0,783	0,7	9'0	0,7	0,703	8'0	0,7	0,778	0,735	0,7	0,7	0,7918	0,7	0,801	0,7381	0,777
$ ho_{ m coh}^{sum}$	187,32	202,14	231,33	213,97	186,9	172,96	224,86	184,99	185,82	197,53	217,07	178,71	194,64	177,19	173,4	221,01	194,65	172,72	209,03
$(Q_{ m AREA}^{\mu})$	158,7	186,85	176,03	182,51	181,55	212,88	173,13	188,08	151,12	166,44	160,78	205,26	193,09	211,82	164,42	160,39	168,83	190,32	187,45
$Q_{ m MIN}^{ocl}$	0,687 5	0,59459	0,65217	89'0	0,476 19	0,39474	0,789 47	0,41176	0,714 29	0,5082	0,597 01	0,441 18	0,472 22	0,371 43	1,0	0,555 56	0,622 22	0,58333	0,652 17
Q _{MIN}	0,5	0,405 41	0,173 91	0,2	0,309 52	0,5	0,421 05	0,205 88	0,285 71	0,2459	0,223 88	0,470 59	0,111111	0,11429	0,125	0,2963	0,244 44	0,444 44	0,173 91
$Q_{ m MIN}^{com}$	12	32	15	21	36	29	14	31 (18	45	46	28	32	33	8	40	37 (27 (20
$Q_{ m MIN}^{cnt}$	16	37	23	25	42	38	19	34	21	61	29	34	36	35	8	54	45	36	23
$Q_{ m MMB}$	180,99	198,7	187,51	201,16	194,88	220,08	188,23	200,45	173,21	180,27	176,79	216,15	206,18	224,54	190,02	175,89	184,09	202,17	205,25
Q _{MU}	178,11	196,99	185,06	197,93	193,02	218,62	185,09	197,7	169,5	177,5	173,43	214,18	203,83	222,33	186,79	172,68	181,8	200,11	201,76
$Q_{ m OFL}^{\sigma}$	0,348	0,240	0,280	0,252	0,242	0,263	0,280	0,2317	0,273	0,256	0,215	0,257	0,251	0,2729	0,274	0,373	0,184	0,248	0,213
$(Q_{ m OFL}^{\mu})$	0,212 63	0,1657	0,16953	0,160 46	0,17477	0,17838	0,19754	0,16601	0,182 97	0,17233	0,14448	0,1673	0,173 19	0,212 31	0,17784	0,279 06	0,141 18	0,16904	0,126 07
$Q_{ m RVU}^{\sigma}$	1,089 9	0,522	0,512	0,639	0,912	0,868	0,732	0,749	0,495	0,731	0,647	0,983	0,994	0,733	0,618	0,973	0,724	0,754	0,538
(Q_{RVU}^{μ})	1,173 7	1,061 1	1,014 4	1,114 2	1,150 4	1,161 6	1,1542	1,138 9	1,078	1,087	1,159	1,198 2	1,231 1	1,163 2	1,180 5	1,196 6	1,122 4	1,091 7	1,120 6
$Q_{\mathrm{FDA}}^{\sigma}$	0,223	0,265	0,182	0,279	0,249	0,242	0,2235	0,280	0,193	0,239	0,278	0,234	0,279	0,297	0,256	0,225	0,269	0,2838	0,275
(Q_{FDA}^{μ})	0,499	0,530	0,489	0,5321	0,447	0,480	0,515	0,533	0,489	0,487	0,531	0,475	0,528	0,509	0,519	0,475	0,506	0,528	0,532
$Q_{ m LCS}^{\sigma}$	0,144	0,242	0,139	0,139	0,331	0,200	0,168	0,267	0,141	0,249	0,222	0,201	0,238	0,2993	0,193	0,254	0,222	0,2182	0,187
$(Q_{\rm LCS}^{\mu})$	0,831	0,758	0,829	0,816	0,630	0,752	0,822	0,718	0,832	0,730	0,747	0,754	0,743	0,670	0,800	0,749	0,753	0,749	0,785
$Q_{ m OCL}^{\sigma}$	0,1178	0,159 58	0,082	0,14693	0,182 28	0,13738	0,089	0,131 76	0,097	0,121 63	0,111 13	0,101,09	0,159 03	0,24182	0,131 03	0,167 41	0,142 59	0,149 42	0,12623
(Q_{OCL}^{μ})	0,840 93	0,801 14	0,869 15	0,8245	0,783 23	0,771 97	0,85287	0,789 73	0,856 63	0,82987	0,838 27	0,807 99	0,801 23	0,720 57	0,820 69	0,807 02	0,828 55	0,780 73	0,830 39
029794-4		70	69	69	89	89	29	29	99	99	99	99	64	64	63	63	62	62	61
Image identifier (FVC2000/ Db1/101_1	FVC2000/ Db1/16_5	FVC2000/ Db1/101_6	FVC2000/ Db1/105_8	FVC2000/ Db1/102_5	FVC2000/ Db1/104_3	FVC2000/ Db1/101_5	FVC2000/ Db1/103_4	FVC2000/ Db1/101_2	FVC2000/ Db1/102_8	FVC2000/ Db1/102_2	FVC2000/ Db1/104_2	FVC2000/ Db1/106_1	FVC2000/ Db1/106_8	FVC2000/ Db1/101_3	FVC2000/ Db1/102_4	FVC2000/ Db1/102_7	FVC2000/ Db1/104_5	FVC2000/ Db1/105_6

Table A.1 (continued)

	44	8 4	4 ×	7.3	2 2	0 %	2 2	2 2	7 7	8 2	3 6	8 1	612	3 1	1 0	4 0	0 4	10 vo	1
Q_{COH}^{rel}	0,754 94	0,717 88	0,784 53	0,763	0,692	0,630	0,775	0,785	0,691	0,818	0,726 63	0,781	0,719 95	0,701	0,709	0,664	0,710	0,455	0,696
Q_{COH}^{sum}	186,47	206,03	204,76	206,79	144,82	139,24	194,62	190,12	168,66	183,44	201,28	207,98	143,99	152,17	139,71	163,48	164,75	164,49	192,91
$(Q^{\mu}_{ m AREA})$	218,33	145,91	195,98	178,26	186,83	202,06	180,91	173,58	186,74	163,18	181,57	161,2	221,27	213,47	224,2	155,52	151,38	208,43	188,97
$Q_{\rm MIN}^{ocl}$	0,59091	0,54839	0,857 14	0,58824	0,46667	0,416 67	0,63636	0,513 51	0,5	0,7619	0,44737	0,53191	0,857 14	0,83333	0,65	0,53846	0,46296	0,666 67	0,52174
Q _{MIN}	0,272 73	0,22581	0,14286	0,35294	0,133 33	0,0	0,318 18	0,162 16	0,39286	0,190 48	0,23684	0,2766	0,285 71	0,25	0,4	0,46154	0,166 67	0,666 67	0,3913
Q com	22	24	13	41	28	12	18	34	26	19	31	36	14	12	20	37	46	33	17
$Q_{ m MIN}^{cnt}$	22	31	14	51	30	12	22	37	28	21	38	47	14	12	20	39	54	33	23
Q _{MMB}	228,74	166,32	213,0	189,39	206,54	212,86	194,59	194,2	200,002	181,58	194,16	179,52	231,14	228,05	230,38	177,01	179,52	209,65	202,71
Q _{MU}	227,08	161,17	211,21	187,22	203,84	211,42	191,63	191,29	197,41	179,34	191,11	177,37	230,61	226,14	229,86	173,16	175,12	208,54	199,97
$Q_{ m OFL}^{\sigma}$	0,257	0,176	0,238	0,226	0,200	0,230	0,204	0,210	0,249	0,233	0,228	0,166	0,219	0,253	0,194	0,2693	0,206	0,253	0,241
$(Q_{ m OFL}^{\mu})$	0,162 76	0,11129	0,146 01	0,14004	0,15665	0,19141	0,1271	0,153 52	0,189 54	0,14586	0,171 18	0,1234	0,149 47	0,19974	0,1277	0,164 42	0,132 76	0,23627	0,190 63
$Q_{ m RVU}^{\sigma}$	1,328 7	0,873	0,693	0,651	1,322 5	1,013 5	0,578	0,872	0,749	1,069 1	0,768	1,063 6	0,708	0,922	0,635	0,879	0,837	0,927	0,680
(Q_{RVU}^{μ})	1,192 2	1,198 4	1,127 7	1,111 8	1,336 6	1,216 3	1,079 2	1,212 7	1,082 4	1,228 3	1,096 2	1,239 7	1,1118	1,206 7	1,0544	1,1838	1,253 6	1,224 9	1,104 7
$Q_{\mathrm{FDA}}^{\sigma}$	0,245	0,281	0,267	0,272 7	0,309	0,311	0,247	0,322	0,288	0,260	0,255	0,275	0,258	0,304	0,266	0,288	0,323	0,254	0,267 4
(Q_{FDA}^{μ})	0,459	0,535	0,524	0,534	0,552	0,5738	0,476	0,588	0,552	0,490	0,511 68	0,500	0,474	0,525	0,475	0,524	0,560	0,425	0,564
$Q_{\mathrm{LCS}}^{\sigma}$	0,141	0,138	0,199	0,270	0,302	0,226	0,108	0,292	0,2374	0,236	0,165	0,273	0,201	0,320	0,240	0,3184	0,263	0,235	0,170
(Q_{LCS}^{μ})	0,780 9	0,795	0,785	0,700	0,645	0,701	0,816 84	0,685	0,739	0,756	0,789	0,704	0,748	0,608	0,707	0,618	0,6849	0,684	0,773
$Q_{ m OCL}^{\sigma}$	0,223 12	0,10367	0,19465	0,13493	0,232 19	0,18143	0,13721	0,11281	0,14735	0,12048	0,13433	0,177 47	0,1555	0,296 04	0,14602	0,1814	0,13336	0,195 99	0,180 5
(Q_{OCL}^{μ})	0,7578	0,816 07	66 962'0	0,81131	0,682 69	0,684 03	0,815 47	0,842 43	88 692'0	0,84872	0,804 96	0,791 52	0,755 6	0,63911	0,752 55	0,73689	0,782 22	0,609 63	0,752 68
Q ₂₉₇₉₄₋₄	61	09	09	29	29	28	28	57	57	26	26	55	55	54	54	53	53	52	52
Image identifier	FVC2000/ Db1/106_7	FVC2000/ Db1/105_2	FVC2000/ Db1/105_5	FVC2000/ Db1/102_6	FVC2000/ Db1/106_2	FVC2000/ Db1/107_3	FVC2000/ Db1/15_1	FVC2000/ Db1/102_3	FVC2000/ Db1/103_1	FVC2000/ Db1/10_8	FVC2000/ Db1/103_3	FVC2000/ Db1/102_1	FVC2000/ Db1/107_7	FVC2000/ Db1/106_5	FVC2000/ Db1/107_5	FVC2000/ Db1/109_3	FVC2000/ Db1/109_5	FVC2000/ Db1/100_7	FVC2000/ Db1/103_6

Table A.1 (continued)

_ =	0,754	0,704	0,783	0,731	0,686 38	0,843	0,593	0,680	099'0	0,656	0,629	0,761	5 4	0,583	0,676	0,692	0,500	0,691	8 0
Q _{COH}													0,8154						8 099'0
$Q_{\rm COH}^{sum}$	177,41 168,31	153,6	213,17	154,41	115,31	176,22	186,47	153,17	108,4	150,36	123,98	125,59	153,29	115,45	135,91	153,74	76,081	768,86	128,86
$(Q_{ m AREA}^{\mu})$	177,41	180,4	181,26	191,96	186,18	171,0	175,75	140,64	169,47	175,02	209,81	184,77	186,32	202,24	154,18	191,11	194,98	194,94	201,42
$Q_{ m MIN}^{ocl}$	6,0	0,25	89'0	0,666 67	0,41667	0,791 67	0,441 18	0,42424	0,310 34	9,0	0,857 14	0,44444	0,73333	0,666 67	0,3913	0,857 14	I	0,333 33	0,5
Q _{MIN}	0,1	0,0625	0,08	0,222 22	0,083	0,20833	0,176 47	0,18182	996	0,0	0,571 43	0,444 44	0,366 67	0,0	0,173 91	0,0	I	0,222 22	0,166 67
Q com	6	32	19	17	12	14	33	28	27	10	7	6	22	9	21	7	0	6	9
$Q_{ m MIN}^{cnt}$	10	32	25	18	12	24	34	99	29	10	7	6	30	9	23	7	0	6	9
Оммв	200,68	198,0	199,19	205,82	206,17	185,17	186,67	173,13	195,64	198,04	219,82	204,68	212,54	213,77	182,4	203,65	209,55	212,9	215,5
Q _{MU}	198,3	196,12	195,53	203,74	204,05	182,91	182,45	168,13	192,87	196,09	218,74	202,53	209,16	212,03	178,68	201,83	208,66	211,22	213,61
$Q_{ m OFL}^{\sigma}$	0,3567	0,2333	0,192	0,234	0,425	0,2111	0,236	0,203	0,276	0,310	0,227	0,265	0,1649	0,249	0,212	0,217	0,224	0,468	0,245
$(Q_{ m OFL}^{\mu})$	0,257 26	0,15833	0,122 68	0,14685	0,447 61	0,12846	0,163 92	0,130 97	0,217 49	0,227 08	0,207 67	0,18891	0,099 617	0,19454	0,16809	0,13504	0,2076	0,502 19	0,22426
$Q_{ m RVU}^{\sigma}$	1,226 4	1,179 3	0,584	0,877	0,808 5	0,820	1,114 2	1,013 5	1,016 7	0,938	1,036 5	1,5533	0,395	0,797	1,141 1	0,797 1	0,920	1,161 6	1,131 1
(Q_{RVU}^{μ})	1,182 9	1,2678	1,106 9	1,257 7	1,090 6	1,130 3	1,330 7	1,282 1	1,223 5	1,177 3	1,2063	1,2698	1,061 9	1,127 8	1,249 7	1,1198	1,214 1	1,221 1	1,237 1
$Q_{\mathrm{FDA}}^{\sigma}$	0,270	0,320	0,279	0,289	0,249	0,241	0,323	0,323	0,326	0,285	0,303	0,2861	0,222	0,324	0,325	0,2941	0,289	0,2713	0,316
(Q_{FDA}^{μ})	0,585	0,660 5	0,544	0,560	0,513	0,476	0,579	0,530	0,601	0,578 6	0,613	0,574	0,472	0,6198	0,620	0,537	0,506	0,529	0,612
$Q_{\mathrm{LCS}}^{\sigma}$	0,226	0,297	0,189	0,139 86	0,207	0,261	0,316	0,304	0,255	0,268	0,264 1	0,160	0,106	0,249	0,283	0,271	0,207	0,224	0,321
(Q_{LCS}^{μ})	0,764	0,669	0,782	0,780	0,747	0,7263	0,607	0,644	0,683	0,715	0,675	0,766	0,8153	0,683	0,660	0,704	969'0	0,738	0,632
$Q_{ m OCL}^{\sigma}$	0,167 78	0,173 73	0,140 71	0,145 48	0,223 61	0,094	0,19537	0,12805	0,188 73	0,20152	0,243 85	0,188 66	0,121 06	0,183 21	0,192 09	0,18539	0,181 63	0,229 43	0,2519
(Q_{OCL}^{μ})	0,770 83	0,756 69	0,827	0,762 27	0,658 65	0,866 92	0,716 42	0,781 17	0,68427	80 602'0	0,633 61	0,734 09	0,836 03	0,639 03	0,717 66	0,739 17	0,573 75	0,641 25	0,641 51
$Q_{29794-4}$		51	20	20	49	49	48	48	47	47	46	46	45	45	44	44	43	43	42
Image identifier (FVC2000/ Db1/101_4	FVC2000/ Db1/108_4	FVC2000/ Db1/105_7	FVC2000/ Db1/107_2	FVC2000/ Db1/12_2	FVC2000/ Db1/30_6	FVC2000/ Db1/108_7	FVC2000/ Db1/109_7	FVC2000/ Db1/108_3	FVC2000/ Db1/14_3	FVC2000/ Db1/16_7	FVC2000/ Db1/22_1	FVC2000/ Db1/15_5	FVC2000/ Db1/21_4	FVC2000/ Db1/108_1	FVC2000/ Db1/11_3	FVC2000/ Db1/110_5	FVC2000/ Db1/12_4	FVC2000/ Db1/13_1

Table A.1 (continued)

	18 59	84 51	9	27	12 93	4 [47	52 75	90	96 13	49 13	58	18 56	98 88	72 38	61 23	63	08 37	2
$Q_{ m COH}^{rel}$	0,618 59	0,684 51	0,7016	0,727	2'0	0,614	0,647	9'0	0,490	0,696	0,649 13	0,558	0,618	0,462 88	0,672	0,661	0,663	0,508	0,6102
Q_{COH}^{sum}	118,77	178,66	137,51	146,9	101,24	121,63	167,59	181,46	89,268	137,14	130,48	69,829	127,42	91,187	147,92	169,27	147,24	85,406	123,87
$(Q^{\mu}_{ m AREA})$	205,91	159,22	180,83	190,73	210,11	199,22	162,35	158,67	204,29	191,6	193,73	193,66	196,61	201,78	182,75	152,21	155,78	194,89	177,41
Q ocl	0,666 67	0,363 64	0,714 29	0,5	0,45455	0,14286	0,44186	0,28889	I	1,0	0,785 71	0,0	0,46667	I	6'0	1,0	0,275 86	0,0	0,227 27
Q _{MIN}	0,333 33	0,272 73	0,285 71	0,428 57	0,0	0,035	0,139 53	0,244 44	I	0,25	0,0	0,0	0,0	I	0,2	0,833 33	0,068	0,0	0,045
Q com	9	21	34	14	11	28	41	40	0	12	14	2	15	0	10	9	27	2	42
$Q_{ m MIN}^{cnt}$	9	22	35	14	11	28	43	45	0	12	14	2	15	0	10	9	29	2	44
Q_{MMB}	219,74	180,16	194,76	207,3	232,75	216,75	185,61	180,2	211,82	212,54	213,04	206,55	210,9	207,18	207,48	177,62	185,88	208,41	196,13
Q _{MU}	218,45	176,8	192,42	204,82	231,86	214,92	180,11	175,25	211,92	209,64	210,53	205,49	208,86	206,24	203,92	173,62	181,18	207,06	194,16
$Q_{ m OFL}^{\sigma}$	0,200	0,254	0,213	0,259	0,317	0,261	0,162	0,261	0,205	0,272	0,230	0,224	0,318	0,238	0,313	0,199	0,391	0,353	0,291
$(Q_{ m OFL}^{\mu})$	0,170 99	0,18467	0,133 48	0,22827	0,27283	0,226 69	0,10025	0,16951	0,185 78	0,255 53	0,147 54	0,20888	0,28636	0,253 69	0,26474	0,118 07	0,310 71	0,336 68	0,223 11
$Q_{ m RVU}^{\sigma}$	1,595 9	1,123 5	0,608	1,836 2	0,974	1,568 1	1,335 6	1,071 4	1,096 5	1,117 5	1,309	0,930	1,627 5	1,0528	1,056 6	1,299 3	0,884	1,218 3	1,0688
(Q_{RVU}^{μ})	1,333 6	1,263 4	1,1168	1,4268	1,198	1,3418	1,3631	1,177 6	1,138 1	1,282 3	1,275 1	1,208 7	1,459 6	1,256	1,191 7	1,3814	1,121 7	1,3687	0,320 1,228 9
$Q_{\mathrm{FDA}}^{\sigma}$	0,322	0,287	0,263	0,317	0,3193	0,325	0,306	0,324	0,309	0,327	0,331	0,3186	0,309	0,289	0,321	0,338	0,2901	0,287	0,320
(Q_{FDA}^{μ})	0,603	0,550	0,477	0,584	0,526	0,597	0,504	0,627 7	0,481	0,613	0,632	0,514	0,636	0,497	0,536	0,654	0,569	0,560	0,626
$Q_{\mathrm{LCS}}^{\sigma}$	0,319	0,264	0,178	0,257	0,271	0,295	0,223	0,287	0,253	0,346	0,326	0,272	0,227 4	0,191	0,252	0,327	0,2515	0,240	0,275
(Q_{LCS}^{μ})	0,587	0,717	0,750	0,693	0,667	0,603	0,693	0,655	0,655	0,575	0,597	0,634	0,708	0,702	0,680	0,618	0,703	0,699	0,676
$Q_{ m OCL}^{\sigma}$	0,229 99	0,160 09	0,097	0,275 49	0,22581	0,23111	0,12626	0,187 27	0,18639	0,244 65	0,203 46	0,169 09	0,2417	0,133 62	0,247 21	0,15495	0,20874	0,15999	0,21471
(Q_{OCL}^{μ})	0,622 48 0,229 99	0,7615	0,779 03	0,655 53	0,6511	0,626 92	0,742 83	0,715 41	86 692'0	0,671	0,670 31 0,203 46	0,61104	0,623 88	0,59492	0,665 07	0,757 47	0,673 38	0,596 94	0,636 16
Q ₂₉₇₉₄₋₄	42	41	41	40	40	39	39	38	38	37	37	36	36	35	35	34	34	33	33
Image identifier	FVC2000/ Db1/20_3	FVC2000/ Db1/61_2	FVC2000/ Db1/67_3	FVC2000/ Db1/4_1	FVC2000/ Db1/42_6	FVC2000/ Db1/75_7	FVC2000/ Db1/79_8	FVC2000/ Db1/108_8	FVC2000/ Db1/110_7	FVC2000/ Db1/64_7	FVC2000/ Db1/75_4	FVC2000/ Db1/110_8	FVC2000/ Db1/27_1	FVC2000/ Db1/100_5	FVC2000/ Db1/76_4	FVC2000/ Db1/109_2	FVC2000/ Db1/50_8	FVC2000/ Db1/100_8	FVC2000/ Db1/108_2

Table A.1 (continued)

	$arrho_{ ext{COH}}^{rel}$	0,570	0,611	0,602	0,618	0,589	0,593	0,733	0,599	0,702	0,501	0,505	0,360	0,643	0,573	0,588	0,571	0,551	0,571	0,555
	$ ho_{ m coh}^{sum}$	132,3	106,46	126,48	107,68	136,77	152,46	88,709	133,11	113,03	204,59	158,85	301,86	119,6	113,02	115,98	82,903	81,021	116,11	90,029
	$(Q_{ m AREA}^{\mu})$	151,59	179,1	195,28	185,92	164,94	153,6	184,24	159,88	198,13	120,26	148,07	167,14	174,01	189,58	187,36	170,53	170,69	186,98	166,37
	$Q_{ m MIN}^{ocl}$	0,333 33	9,0	0,769 23	0,25	0,088	0,044	0,666 67	0,21818	0,4	0,022	0,25	0,466 67	0,75	0,095	1,0	0,375	0,058	0,107 14	0,428 57
	$Q_{ m MIN}^{ m mu}$	0,066	0,0	0,076	0,0	0,058	0,10294	0,0	0,12727	0,45	0,229 89	0,5	0,166 67	0,5	0,095	0,0	0,125	0,058	0,0	0,14286
	$Q_{ m MIN}^{com}$	14	2	13	8	34	89	9	52	20	54	8	38	4	21	3	8	17	28	7
	$Q_{ m MIN}^{cnt}$	15	5	13	8	34	89	9	52	20	87	8	09	4	21	3	8	17	28	7
	Q_{MMB}	179,32	196,84	214,74	210,65	189,99	177,44	213,46	188,16	219,22	154,45	161,39	167,11	201,59	210,81	212,6	202,75	194,32	211,16	197,47
	$ ho_{ ext{MU}}$	176,33	194,98	212,49	208,02	187,8	174,26	211,68	182,95	217,13	154,63	157,83	167,2	198,32	208,17	209,28	200,07	193,06	208,54	194,37
	$arrho_{ m OFL}^{\sigma}$	0,215 88	0,184	0,210 85	0,238	0,270	0,228	0,402	0,184 84	0,392	0,454	0,203	0,467	0,268	0,222	0,363	0,2359	0,242	0,226	0,2385
	$(Q_{ m OFL}^{\mu})$	0,19288	0,14262	0,17781	0,17922	0,21338	0,175 43	0,35845	0,127 59	0,34448	0,53997	0,195 58	0,623 49	0,2031	0,14925	0,329 41	0,17834	0,19045	0,17878	0,17887
	$Q_{ m RVU}^{\sigma}$	1,1358	1,286	1,587 6	1,1083	1,220 7	0,971	0,959 5	1,106 6	1,172 2	1,3462	1,233 9	1,6582	1,6898	1,605 9	1,5893	1,244 5	1,2049	1,760 1	2,027 2
	(Q_{RVU}^{μ})	1,250 6	1,321 6	1,409 4	1,2645	1,236 5	1,172 3	1,242 1	1,305 2	1,311 6	1,31	1,331 4	1,475 2	1,450 3	1,445	1,341 6	1,3419	1,387 6	1,356 5	1,680 2
	$Q_{ ext{FDA}}^{\sigma}$	0,327	0,336	0,337	0,351	0,336	0,323 5	0,327	0,327	0,317	0,327	0,313	0,359	0,353	0,346	0,3432	0,344	0,369	0,353	0,342
	(Q_{FDA}^{μ})	0,638	0,547	0,626	0,717	0,663	0,634	0,582	0,6044	0,567	0,568	0,6189	0,646	0,672	0,695	0,597	0,610	0,632	0,6278	0,696
	$Q_{ m LCS}^{\sigma}$	0,201	0,257	0,345	0,292	0,263	0,228	0,2561	0,191	0,299	0,372	0,250	0,382	0,2845	0,352	0,304	0,243	0,3028	0,342	0,301
	(Q_{LCS}^{μ})	0,717	0,631	0,552	0,614	0,677	96960	0,669	60,700	0,616	0,461	0,694	0,376	0,616	0,495	0,595	0,630	0,559	0,520	0,575
	$Q_{ m OCL}^{\sigma}$	0,18737	0,23155	0,23117	0,142 42	0,16532	0,15538	0,29648	0,17133	0,243 03	0,23427	0,19479	0,25032	0,26771	0,23073	0,25845	0,205 99	0,207 55	0,22619	0,207 68
	(Q_{OCL}^{μ})	0,666 43	0,589 18	0,635 26	0,677 77	0,692 54	0,71695	0,54088	0,6784	0,626 28	0,465 67	0,622 96	0,47433	0,59485	0,5832	0,588 17	0,5735	0,57831	0,602 68	0,57787
	0 29794-4		32	31	31	30	30	59	56	28	28	27	27	26	26	25	25	24	24	23
Ітада	ı.	FVC2000/ Db1/78_3	FVC2000/ Db1/91_1	FVC2000/ Db1/75_2	FVC2000/ Db1/89_3	FVC2000/ Db1/78_2	FVC2000/ Db1/78_4	FVC2000/ Db1/48_3	FVC2000/ Db1/70_6	FVC2000/ Db1/85_7	FVC2000/ Db3/100_2	FVC2000/ Db1/84_4	FVC2000/ Db3/109_1	FVC2000/ Db1/86_8	FVC2000/ Db1/94_7	FVC2000/ Db1/76_2	FVC2000/ Db1/87_7	FVC2000/ Db1/93_5	FVC2000/ Db1/94_1	FVC2000/ Db1/87_5

Table A.1 (continued)

		2 +	3 0	m m	L 1	4 +	0 %	3 0	4	3.5	1	7.5	5.7	2.5	4 10	4	1 8	æ ++	_
$Q_{ m COH}^{rel}$	0,587	0,452	0,549 23	0,413	0,515	0,484	0,370	0,369	0,2734	0,412	0,639 51	0,642	0,337	0,522 96	0,484	0,584	0,548	0,518	0,6137
Q_{COH}^{sum}	121,06	8'96	101,06	148,48	81,903	70,316	87,11	177,65	108,27	148,98	113,83	123,33	180,49	101,98	6,87	112,13	77,903	153,34	170,88 111,69
$(Q^{\mu}_{ m AREA})$	189,11	202,76	137,14	208,22	162,7	194,52	206,8	186,77	179,26	211,18	168,15	180,53	183,69	202,19	163,46	187,95	162,8	146,77	170,88
$Q_{ m MIN}^{ocl}$	0,135 14	I	0,2	I	0,44444	I	I	0,0	0'0	I	0,583 33	0,84615	0,0	0,074	1,0	0,03125	0,19048	0,25	0,14286
Q _{MIN}	0,081	I	0,0	I	0,0	I	I	0,68153	0,34483	I	0,166 67	0,46154	0,35484	0,148 15	1,0	0,1875	0,047	0,375	0,0
$Q_{ m MIN}^{com}$	37	0	2	0	6	0	0	91	96	0	12	13	29	27	Н	32	21	8	7
$Q_{ m MIN}^{cnt}$	37	0	2	0	6	0	0	157	145	0	12	13	62	27	₩	32	21	8	7
$Q_{\rm MMB}$	209,94	209,58	175,65	209,61	198,63	209,67	212,93	208,76	198,97	212,0	194,69	204,57	196,93	218,31	194,16	210,64	195,29	164,02	196,65
Q _{MU}	207,28	208,75	172,12	208,52	195,14	209,23	212,2	208,75	199,17	211,27	192,38	201,51	197,17	216,61	190,32	208,15	192,53	160,19	193,41
$Q_{ m OFL}^{\sigma}$	0,241	0,200	0,161	0,224	0,293	0,253	0,194	0,4206	0,389	0,226	0,317	0,320	0,462	0,229	0,266	0,237	0,187	0,187	0,218
$(Q_{ m OFL}^{\mu})$	0,185 22	0,177 47	0,135 95	0,243 34	0,225 01	0,252 31	0,178 52	0,49887	0,489 33	0,23412	0,302	0,340 32	0,699 5	0,203 66	0,213 56	0,167 21	0,15803	0,18893	0,149 71
$Q_{\mathrm{RVU}}^{\sigma}$	1,262 6	1,0943	1,753 6	1,3008	1,728 4	0,809	1,166 7	1,402 6	1,397	1,522 3	1,7208	1,333 3	1,631 7	1,481 6	1,300 1	1,692 5	1,537 4	1,352 2	1,069 5
$(Q_{\rm RVU}^{\mu})$	1,346	1,2948	1,4401	1,358	1,393 7	1,1258	1,242 6	1,332 9	1,3638	1,477 9	1,416 2	1,262 1	1,381 4	1,418 9	1,384	1,457 7	1,420 1	1,413	1,263 3
$Q_{\mathrm{FDA}}^{\sigma}$	0,346	0,3378	0,355	0,296	0,330	0,3165	0,333	0,2762	0,280	0,320	0,331	0,332	0,3068	0,359	0,329	0,355	0,331	0,312	0,339
(Q_{FDA}^{μ})	0,651	0,569	0,690	0,497 4	0,727	0,541	0,595	0,5092	0,474	0,492	0,610	0,644	0,538	0,615	0,7583	0,690	0,760	0,662	0,682 5
$Q_{\mathrm{LCS}}^{\sigma}$	0,322	0,274	0,271	0,207	0,298	0,202	0,279	0,246	0,275	0,281	0,233	0,289	0,344	0,309	0,286	0,319	0,239	0,239	0,261
(Q ^{\mu_{LCS})}	0,558	0,615	0,624 58	0,708	0,572	0,684	0,600	0,656	0,602	0,6016	0,676	0,6253	0,528	0,550	0,605 9	0,554	0,623	0,692	0,648
$Q_{ m OCL}^{\sigma}$	0,23977	0,16256	0,19982	0,14828	0,17337	0,175 79	0,167 1	0,197 61	0,15642	0,15661	0,246 02	0,262 73	0,15195	0,197 19	0,1903	0,239 92	0,17131	0,20129	0,157 75
(Q_{OCL}^{μ})	0,603 18	0,56621	0,571 92	0,583 06	0,57164	0,5441	0,512 19	0,486 49	0,37131	0,578 38	0,576 67	0,597 92	0,42888	0,567 66	0,586 63	0,590 08	0,58187	0,622 24	0,667 77 0,157 75
029794-4	23	22	22	21	21	20	20	19	19	18	18	17	17	16	16	15	15	14	14
Image identifier	FVC2000/ Db1/94_6	FVC2000/ Db1/92_6	FVC2000/ Db1/93_3	FVC2000/ Db1/100_6	FVC2000/ Db1/87_6	FVC2000/ Db1/110_6	FVC2000/ Db1/92_8	FVC2000/ Db3/104_1	FVC2000/ Db3/105_2	FVC2000/ Db1/110_4	FVC2000/ Db1/91_5	FVC2000/ Db1/76_7	FVC2000/ Db3/15_4	FVC2000/ Db1/97_5	FVC2000/ Db1/99_5	FVC2000/ Db1/94_5	FVC2000/ Db1/95_8	FVC2000/ Db1/84_3	FVC2000/ Db1/89_2

Table A.1 (continued)

	rel COH	0,567	0,512	0,356	0,624	0,443	0,403	0,579	0,448	0,354	0,476	0,462	0,483	0,525	0,565	0,538	0,420	0,345	0,429	0,359
-	0	_		· ·																
	$arrho_{ m COH}^{sum}$	80,037	162,53	44,548	91,208	94,531	54,057	111,92	81,568	78,255	83,839	79,478	88,894	87,243	107,93	118,46	220,56	124,85	87,642	44,181
	$(Q^{\mu}_{ m AREA})$	167,54	165,82	224,31	171,41	202,1	220,44	181,54	206,86	218,41	192,38	186,74	173,36	175,74	160,22	194,61	96,694	203,63	194,47	221,14
	$Q_{ m MIN}^{ocl}$	0,0	0,10638	I	0,0	0,0		0,0	0,333 33		0,0	0,0	0,0	0,0	0,0	0,0	0,005	0,0	0,0	I
	Q _{MIN}	0,0	0,148 94		0,0	0,0		0,05	0,0		0,2	0,25	0,15385	0,0	0,0	0,2	0,324 18	0,25	0,3125	1
	$Q_{ m MIN}^{com}$	9	42 (0	4	1	0	20	3	0	15	4	13 (₩	8	10	110 (8	16	0
	$Q_{ m MIN}^{cnt}$	9	47	0	4	1	0	20	3	0	15	4	13	П	8	10	182	8	16	0
	$Q_{ m MMB}$	191,04	176,8	229,73	198,03	209,06	226,67	208,14	223,78	222,41	209,4	201,85	190,33	202,76	190,29	210,08	147,11	205,58	210,4	227,2
	$Q_{ m MU}$	189,75	172,03	229,17	195,83	208,41	226,33	205,15	222,2	222,26	207,29	200,12	188,95	200,04	187,27	208,0	147,31	203,85	208,71	226,82
	$Q_{ m OFL}^{\sigma}$	0,1929	0,237	0,235	0,332	0,215 28	0,2658	0,189	0,223	0,219 2	0,237	0,277 9	0,262	0,176	0,3059	0,276	0,351	0,261	0,279	0,212 2
	$(Q_{ m OFL}^{\mu})$	0,147 99	0,1754	0,291 31	0,295 17	0,177 48	0,258 62	0,167 17	0,21459	0,226 09	0,208 44	0,254	0,19654	0,172 71	0,265 76	0,245 74	0,362 42	0,2528	0,23851	0,278 52
	$Q_{ m RVU}^{\sigma}$	1,422 6	1,171 1	1,139 7	1,209 5	1,0198	1,321 9	2,108 1	1,478	1,313 2	1,8838	1,707 2	1,256 4	1,026 5	1,005 7	1,178 5	1,199 4	1,720 4	1,333 6	1,72
	(Q^{μ}_{RVU})	1,441 7	1,350 2	1,3002	1,312	1,241	1,342 1	1,5538	1,418 6	1,355 1	1,583 2	1,437 9	1,334 5	1,276 3	1,2687	1,298 4	1,308	1,373 6	1,4103	1,392 4
	$Q_{ ext{FDA}}^{\sigma}$	0,354	0,330	0,3331	0,335	0,357	0,347	0,349	0,344	0,364	0,347	0,339	0,321	0,327	0,3495	0,349	0,305	0,337	0,356	0,362
	(Q_{FDA}^{μ})	0,627	0,684	0,504	0,573	0,615	0,557	0,570	0,719	0,582	0,700	0,714	0,503	0,735	0,613	0,641	0,503	0,6948	0,652 4	0,625
	$Q_{ m LCS}^{\sigma}$	0,311	0,264	0,273	0,2405	0,251	0,231	0,305	0,315	0,296	0,230	0,240	0,279	0,250	0,268	0,334	0,255	0,275	0,265	0,276
	(Q_{LCS}^{μ})	0,553 9	0,654	0,610	0,661	0,620	0,634	0,602	0,510	0,542	0,637	0,624	0,602	0,626	0,630	0,539	0,645	0,593	0,596	0,592
	$Q_{ m OCL}^{\sigma}$	0,20679	0,175 65	0,15847	0,236 48	0,15939	0,16428	0,21494	0,20894	0,178 04	0,19427	0,20694	0,182 72	0,19088	0,248 65	0,22467	0,213 11	0,20134	0,19443	0,140 67
	(Q_{OCL}^{μ})	0,573 44 0,206 79	0,633 18	0,499 98	0,543 66	0,561 12	0,494 41	0,590 59	0,507 78	0,477 29	0,52981	0,491 72	0,57881	0,578 12	0,53437	0,559 08	0,514 53	0,49085	0,527 93	0,484 18
	$Q_{29794-4}$ (13	12	12	11	11	10	10	6	6	8	8	7	7	9	9	2	D	4
	Image identifier	FVC2000/ Db1/93_6	FVC2000/ Db1/95_1	FVC2000/ Db1/100_4	FVC2000/ Db1/91_2	FVC2000/ Db1/92_7	FVC2000/ Db1/98_2	FVC2000/ Db1/97_4	FVC2000/ Db1/97_8	FVC2000/ Db1/110_2	FVC2000/ Db1/90_8	FVC2000/ Db1/90_6	FVC2000/ Db1/91_7	FVC2000/ Db1/89_4	FVC2000/ Db1/93_2	FVC2000/ Db1/96_8	FVC2000/ Db3/102_4	FVC2000/ Db1/90_5	FVC2000/ Db1/90_7	FVC2000/ Db1/100_2

Table A.1 (continued)

$Q_{ m COH}^{rel}$	0,344	0,327	0,4218	0,319	0,294	0,386	0,315	0,275
		43,91						
$\left \left(Q_{ m AREA}^{\mu} ight) ight Q_{ m COH}^{sum}$	228,49 81,965	221,53	203,49 76,346	229,52 86,013	190,82 110,96	126,39 206,14	181,95 160,37	132,59 142,55
$Q_{\rm MIN}^{ocl}$ (I		I	0,0	0,004	0,0	0,0
Q _{MIN}				I	134 0,477 68	122 0,319 15	0,428 57	121 0,450 98
$Q_{ m MIN}^{com}$	0	0	0	0	134	122	84	121
$Q_{ m MIN}^{cnt}$	0	0	0	0	224	235	203	255
$Q_{\rm MMB}$	232,65	224,41	220,09	230,73	207,21	164,54	192,3	146,77
$Q_{\rm MU} \mid Q_{\rm MMB} \mid$	232,05	0,246 224,46 76	218,48	0,247 230,37 07	0,417 207,42 93	164,66	0,411 192,48 98	0,383 146,85 51
$Q_{ m OFL}^{\sigma}$	0,272	0,246	0,240 6 218,48 220,09	0,247	0,417	0,4092 164,66	0,411	0,383
$(Q_{ m OFL}^{\mu})$	0,32912	0,291 09	0,215 04	0,323 82	0,690 01	0,438 07	0,580 75	0,52475
$Q_{ m RVU}^{\sigma}$	1,824 4	1,602 5	1,277 6	1,118 9	1,641	1,181 7	1,658 7	1,3038
$Q^{\mu}_{ m RVU}$)	0,346 1,461 4),333 1,337 7 34),348 1,2945 84),350 1,368 5 95),315 1,487 6 56	1,2862),332 1,557 6 72	,342 1,355 8 34
$\left \begin{array}{c} Q_{\mathrm{FDA}}^{\sigma} \end{array}\right $	0,346	0,333	0,348	0,350	0,315	0,308	0,332	0,342
(Q_{FDA}^{μ})	0,556	0,620	0,675	0,597	0,522	0,515	0,554	0,557
$Q_{ m LCS}^{\sigma}$	0,262	0,303	0,337	0,301	0,332	0,269	0,326	0,315
$(Q_{\rm LCS}^{\mu})$	0,612	0,560	0,470	0,535	0,485	0,629	0,496	0,506
$Q_{ m OCL}^{\sigma}$	0,179 78	0,15134	0,192 27	0,182 15	0,151 1	0,202 45	0,392 0,16715	0,1636
(Q_{OCL}^{μ})	4 0,472 25 0,179 78	0,459 76 0,151 34	0,49986 0,19227	0,432 32 0,182 15	0,360 18	1 0,500 11 0,202 45	0,392	0 0,348 23
$Q_{29794-4} \left[\left(Q_{\mathrm{OCL}}^{\mu} \right) \middle Q_{\mathrm{OCL}}^{\sigma} \middle \left(Q_{\mathrm{LCS}}^{\mu} \right) \middle $	4	3	3	2	2	П	Н	0
Image identifier	FVC2000/ Db1/100_3	FVC2000/ Db1/100_1	FVC2000/ Db1/97_7	FVC2000/ Db1/110_1	FVC2000/ Db3/18_3	FVC2000/ Db3/102_3	FVC2000/ Db3/19_4	FVC2000/ Db3/29_8

Table A.1 (continued)

Image identifier		$\left (Q_{ m OCL}^{\mu}) \right $	$Q_{29794-4} \left \left(Q_{\mathrm{OCL}}^{\mu} \right) \right \left Q_{\mathrm{OCL}}^{\sigma} \right \left(Q_{\mathrm{LCS}}^{\mu} \right) \right $	$\left \left(Q_{\mathrm{LCS}}^{\mu} \right) \right $	$Q_{ m LCS}^{\sigma} \left \left(Q_{ m FDA}^{\mu} ight) ight.$		$Q_{\mathrm{FDA}}^{\sigma} \mid ($	(Q_{RVU}^{μ})	$Q_{ m RVU}^{\sigma} \ \Big \ ($	$(Q_{ m OFL}^{\mu})$	$Q_{ m OFL}^{\sigma} \mid \zeta$	$Q_{MU} \mid Q$	$Q_{\rm MMB} \mid Q$	$Q_{\rm MIN}^{cnt} \mid Q_{ m j}$	$Q_{\rm MIN}^{com} \mid Q_{ m J}$	$Q_{\rm MIN}^{\rm mu}$	$Q_{\rm MIN}^{ocl}$ ($(Q_{ m AREA}^{\mu})$	$Q_{ m COH}^{sum}$	$Q_{ m COH}^{rel}$
FVC2000/ Db3/30_5		0 0,382 67	0,382 67 0,156 85	0,461	0,330	0,538 36	0,340 1,395 9 76		1,380 2	0,50135	$\begin{vmatrix} 0,405 \\ 51 \end{vmatrix}$		145,89	247	111 0,5	0,502 02	0,0	129,92 152,99	152,99	0,303
Key																				
029794-	$Q_{29794-4}$: Unified quality score	quality sco	re																	
$Q_{ m OCL}^{\mu}: m Me$	$Q_{ exttt{OCL}}^{\mu}$: Mean of local orientation certainty level	vrientation	certainty le	evel																
$Q_{ m OCL}^{\sigma}:{ m St}$	$Q^{oldsymbol{\sigma}}_{ ext{OCL}}$: Standard deviation of local orientation certainty level	ation of lo	cal orientati	ion certain	ıty level															
$Q^{\mu}_{ m LCS}$: Me.	$Q^{\mu}_{ extsf{LCS}}$: Mean of local clarity score	larity scor	a																	
$Q_{\mathrm{LCS}}^{\sigma}: \mathrm{St}_{\epsilon}$	$q_{ t LCS}^{\sigma}$: Standard deviation of local clarity score	ation of loc	al clarity sc	core																
$Q_{\mathrm{FDA}}^{\mu}: M\epsilon$	$q_{ ext{FDA}}^{\mu}$: Mean of local frequency domain analysis	requency	domain ana	lysis																
$Q_{\mathrm{FDA}}^{\sigma}:S^{\mathrm{t}}$	$Q_{ ext{FDA}}^{oldsymbol{\sigma}}$: Standard deviation of local frequency domain analysis	iation of lo	cal frequen	cy domain	analysis															
$Q^{\mu}_{\mathrm{RVU}}: M\epsilon$	$Q^\mu_{ ext{RVU}}$: Mean of local ridge valley uniformity	ridge valle	y uniformity	y																
$Q_{\mathrm{RVU}}^{\sigma}$ st	$Q_{ m RVU}^{\sigma}$: Standard deviation of local ridge valley uniformity	iation of lo	cal ridge va	ılley unifor	mity															
$Q^{\mu}_{ m OFL}$: Me	$Q_{ m 0FL}^{\mu}$: Mean of local orientation flow	rientation	flow																	
$\left Q_{ m OFL}^{\sigma} : { m St.} ight $	$Q_{ m OFL}^{\sigma}$: Standard deviation of orientation flow	iation of o	rientation f.	low																
$Q_{MU}: MU$	Ţ.																			
$Q_{\text{MMB}}: MMB$	MMB																			
Q_{MIN}^{cnt} : M	$Q_{ m MIN}^{cnt}$: Minutiae count	nt																		
$Q_{\text{MIN}}^{com}: M$	$q_{ m MIN}^{com}$: Minutiae count in centre of mass	nt in centr	e of mass																	
$Q_{MIN}^{mu}:M$	$ ho_{ extsf{MIN}}^{ extsf{mu}}$: Minutiae quality based in image mean	lity based	in image mo	ean																
Q_{MIN}^{ocl} : M	$Q_{ m MIN}^{\it ocl}$: Minutiae quality based on orientation certainty level	lity based	on orientat.	ion certair	ıty level															
$Q_{AREA}^{\mu}: F$	$arrho_{ ext{AREA}}^{\mu}$: Region of interest image mean	erest imag	ge mean																	
$Q_{\text{COH}}^{sum}: \mathbb{R}_{\epsilon}$	Q^{sum}_{COH} : Region of interest orientation map coherence sum	erest orien	tation map	coherence	mns															
$\left egin{array}{c} Q_{ ext{COH}}^{rel} : ext{R}_{\epsilon} \end{array} ight $	$Q_{ m COH}^{rel}$: Region of interest relative orientation map coherence sum	est relati	ve orientat	ion map cc	sherence s	'nm														

Annex B

(informative)

Factors influencing fingerprint image character

B.1 General

A finger image obtained from a scanner is not always of good utility. It may contain poor character area caused by the user character (e.g. user's skin condition), user behaviour (e.g. improper finger placement), imaging (e.g. scanner limitation or imperfection), or environment (e.g. impurities on the scanner surface).

The performance of an automated fingerprint recognition system will be affected by the amount of poor character area or the degree of artefact present in the finger image. Computing the quality score of the fingerprint image produced is necessary as it should be used by the system to perform quality based recommendation or filtering or adjustment. For example, an enrolment application may recommend performing a second acquisition of a fingerprint when the quality of the first acquisition is below a certain value. Clause 5 defines approaches to compute the finger image quality. The quality score should be predictive of the performance of an automatic fingerprint recognition system.

B.2 Defect factors

This subclause gives some detailed examples of quality metrics impact on a defect.

A fingerprint sensor may produce a fingerprint image with poor quality when a finger is too wet, for example, an outdoor fingerprint access control terminal after rain. For a TIR (Total Internal Reflection) sensor, the presence of water on the sensor induces reflection of the lighting in the water which interferes with the reflection of the lighting on the finger ridges. The image provided may be completely dark in the water area, or with very low contrast as the valleys and ridges gray-level values will be very close.

For such images, the quality metrics proposed in <u>5.2</u> and <u>5.3</u> will return a low value as the distinction of the valleys and ridges will not be done by the algorithm, for example, ridge valley uniformity quality. Especially when the image is composed of almost 2 regions with no contrast, the background and the wet area.

Some defects and their factors can be listed as follows:

- a) Defect caused by user character
 - 1) extreme skin conditions such as very wet, very dry, etc.
 - 2) scars
 - 3) wrinkles
 - 4) blisters
 - 5) eczema
 - 6) impurities such as dirt, latent print, etc.
- b) Defect caused by imaging
 - 1) sampling error
 - 2) low contrast or signal-to-noise ratio

- 3) distortion
- 4) erroneous or streak lines
- 5) uneven background
- 6) insufficient dynamic range
- 7) non-linear or non-uniform grayscale output
- 8) pixels not available due to hardware failure
- 9) aliasing problems
- c) Defect caused by user behaviour
 - 1) elastic deformation
 - 2) improper finger placement such as too low, rotated, etc.
 - 3) insufficient area of finger image
- d) Defect caused by environment
 - 1) humidity
 - 2) light
 - 3) Impurities on the scanner surface such as latent prints

Annex C (informative)

Area consideration

Even though area alone shall not be used as an indicator of quality, the area in the image containing valid fingerprint patterns will affect the performance of the matcher and thus, the quality of the image. If the valid area is too small, then the finger image shall be considered poor. The local analysis operates on a local region basis while the global analysis operates on the entire image without consideration for the actual area occupied by the ridge-valley structure. In the global analysis, there is a possibility that a small but very good quality region may produce sufficient overall score to give the entire image an acceptable quality. Therefore, an explicit area measure is needed. The finger area is calculated based on the number of local regions occupied by the valid ridge-valley structure (the foreground) of a finger image. The background, sweat or other impurities present in the image but not located within the valid ridge-valley structure of the finger image shall be excluded from the calculation of the finger area. The border region of the image may also be excluded.

The process of automatically classifying the image into the foreground which contains valid ridge-valley structure and the remainder as background is known as fingerprint segmentation. One of the segmentation approaches is to compute the strength of the orientation in each local region^[1]. From the segmentation process, the number of local regions, or in general, the area in the foreground and background of the fingerprint can be known, thus providing a possible area consideration in the overall quality of the image.

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