The main goal of this paper is to develop classification of non-contact respiration monitoring approaches and proposal of structure for system with facial artifacts rejection. All available techniques were divided into two main groups: based on reconstruction of respiration from 3-D image of object and based on 2-D image processing of techniques. Structure of system for respiration monitoring using optical sensors with facial artifacts removing was developed. New approach allows improving of respiration monitoring for objects in supine position and in a sitting position. References 26, figures 16.

Keywords: non-contact measurements, non-contact respiration monitoring, respiration monitoring, breathing monitoring, mimic artefacts, optical flow.

1. General approaches to respiration monitoring in optical domain

If detection in optical domain is used for retrieval of respiration parameters, the movements of human body are usually thought to be directly connected to respiration. To detect respiration, the system should be designed to reconstruct body surface movement from the video sequence, and to estimate the magnitude of surface excursions in real time. Techniques of respiration monitoring differ by the general approach, which is used to extract movements.

In this paper, we propose to divide all optical methods into two groups depending on the dimensionality of captured data used for respiration extraction (Fig. 1).

In the first approach, reconstruction of respiration from 3-D image of object is used. It requires a specialized hardware, such as stereoscopic cameras and/or additional lightening equipment. This approach can be subdivided into two classes, first one using stereoscopic imaging, the second one based on projection methods.

Second approach is based on 2-D image processing techniques with using a single optical sensor, and can be subdivided into optical flow methods and frame subtraction methods.


2. 3-D reconstruction based methods

Many kinds of systems for 3-D objects scanning are available in the market [6]. The taxonomy of optical 3-D scanners is given in Fig 2. In biomedical applications it is reasonable to use non-contact systems based on reflective principles. This approach provides patient's comfort and has no influence on his conditions. Two techniques of 3-D reconstruction are employed for respiration monitoring in most cases. Passive stereoscopic method based only on acquisition of reflected light from external sources, and active method called an active triangulation based on projection of specific markers on object before acquisition.

Fig. 2. An optical 3-D scanner taxonomy [6]

2.1. Stereoscopic methods of respiratory monitoring

Stereoscopic system contains two or more cameras that capture video of the same scene from different locations. It is possible to estimate distance to the object using the system demonstrated in Fig. 3. Depth of 3-D image is directly proportional to focal length of camera lens

\[ \text{depth} = \frac{f b d}{f b d} \]

and inversely proportional to disparity of the same object point in pixels

\[ \text{depth} = \frac{f b d}{f b d} \]

The points of object for different images of the same scene should be robustly identified for correct 3-D reconstruction.

Fig. 3. Stereoscopic system
2.2. Projection-based systems for respiration monitoring

The main difference of active systems is the usage of light emitter to create auxiliary projection markers on object surface (Fig. 4). 3-D position of spot on the object’s surface can be computed as intersection between $l_1$ and $l_2$ by triangulation [6].

Fig. 4. Active triangle system [6]

2.3. Respiratory monitoring using 3-D techniques

Li et al. [3] used stereoscopic camera manufactured by Xigen for respiration tracking tasks in radiotherapy applications. Authors calculated 2-D depth image from 3-D reconstructed surface. After conversion of 3-D surface images to 2-D depth images, they used principal component analysis (PCA) to perform an unsupervised learning to extract different surface motion patterns from the 2-D depth image sequence. For the surface movements’ patterns classification the support vector machine (SVM) technique is used (Fig. 5). Experiments have been conducted on phantom and 4 volunteers.

Fig. 5. The surface of the phantom (left) and the average first eigenvector motion with time (right) [3]

The majority of researchers [5], [7]–[9] uses active 3-D systems based on single CCD camera and fiber grating device (FGD). The FG device generate field of dots or stripes for three-dimensional surface of object reconstruction Fig. 7.

Fig. 7. Fiber grating device and dot-pattern [9]

FG devices are available on the market, Aoki et al. [8] used solution manufactured by Sumitomo Osaka Cement Co, ltd.

Appearance of common construction of experimental system shown in Fig. 8. System included bed with overhead support, where CCD camera and FGD are installed. In the most cases, resolution of CCD cameras was equal to 640x480 pixels.

Fig. 8. Optical scheme of active triangle respiration monitoring system [8]
The relationship between vertical motion, represented by a symbol $\Delta Z$, in the thoracicoabdominal part and shift length of the pattern light, denoted by a symbol $d$, in the image is given by the following equation:

$$\Delta \delta = \frac{\Delta Z}{Z(Z - \Delta Z)} dl$$

where $Z$ is the vertical distance from the height, at which the sensor was installed, to the body surface, $d$ is the focal distance of lens, and $l$ is the horizontal distance from the center of lens to the FG element. Thus the measurement principle of this method is based on the triangulation [5].

Aoki et al. [5], [8] used active triangle system for respiration parameters obtaining (Fig. 6). Authors have shown that waveforms measured by spirometer correspond to the respiratory flow measured by non-contact method. The same authors achieved better results with stripe projections then in the case of dot field projections [8]. They also obtained similar results with and without top covered quilt.

Povsic et al. [10] proposed non-contact real-time system for teaching and correcting respiration. Tamagawa et al. [9] presented system based on active triangle system for respiration estimation during tomography examination.

The Kinect system (9) for 3-D reconstruction is prevalent as well. Kinect was initially developed as gaming device but this system is used in respiratory applications [11]–[13]. Kinect includes RGB camera and couple of IR projector and IR camera, which are used for estimation of depth distance. System captures depth map of image with 320x240 pixel resolution. Yu et al. [14] developed Kinect-based system for monitoring position, body movements, and breathing rate during sleep.

Lodovico et al. [15] and Povsic et al. [10] proposed complex systems for whole human trunk surface visualization during respiration analysis.

Systems in Figures 10 and 11 allow calculation of not only respiratory rate but also of expiratory volume.

3. 2-D reconstruction based methods

Systems with 2-D reconstruction of respiration use single camera to capture video from the object, and consequent processing of two-dimensional images for localization of the object points.

2-D techniques employ single optical sensor for image acquisition. Sensors with various wavelength range can be used. Visible range cameras and near infrared cameras are used usually for respiration monitoring. Both types of cameras can be manufactured by CCD or CMOS technology. Modern CCD and CMOS sensors have comparable characteristics but CMOS is inexpensive and has lower power consumption. Therefore, CMOS cameras are usually used for general purpose and consumer electronics (web-cams, smartphones, etc.) and CCD are used in scientific and industrial applications.

Visible range cameras for respiration monitoring with good scene illumination can be employed. In the work [16], a system of NIR camera and NIR projector are implemented for the respiration monitoring in the scene with low brightness (monitoring during night). Fig. 12 shows sample frame of night respiration monitoring of patient in supine position.
3.1. Respiration extraction by optical flow estimation

The most widespread technique that implements object movement estimation in video sequences is based on assessment of optical flow coming from the patient. For optical flow estimation, a few assumptions have to be accepted. First is that intensity of pixel for the same object on the video does not change between consecutive frames (1). This means that one gets the same intensity of the pixel on the next frame and this pixel is only shifted by \( u_x \) and \( u_y \) distance in \( x \) and \( y \) direction respectively.

\[
I(x,y,t) = I(x + u_x, y + u_y, t + 1) \tag{1}
\]

Consider pixel \( I(x,y,t+1) \) and its copy \( I(x,y,t) \) displaced by \( u \), where \( u = [u_x, u_y] \) is two dimensional magnitude of displacement value (2).

\[
I(x,y,t+1) = I(x,y,t) + u \nabla I(x,y,t) \tag{2}
\]

\[
I(x,y,t) - I(x,y,t+1) + u \nabla I(x,y,t) = 0 \tag{3}
\]

\[
\frac{\partial I(x,y,t)}{\partial t} + u \nabla I(x,y,t) = 0 \tag{4}
\]

Expression (3) has been obtained using Taylor series expansion and ignoring high-order terms. Since only one time instance lies between two consecutive frames, the \( I(x,y,t) - I(x,y,t+1) \) part of equation (3) can be regarded as time derivative \( \frac{\partial I(x,y,t)}{\partial t} \) of pixel intensity for frame \( I(x,y,t) \). Equation (5) was obtained after removing the gradient expression in (4). Equation (5) is called 2-D Motion Constraint Equation or Gradient Constraint. Motion Constraint Equation has no solutions because we have one equation and two unknowns. For solving this equation, one should assume that neighboring pixels have the same shift and consequently have the same magnitude of displacement \( u \).

Having this assumption accepted, it is possible to solve system of equation (6) where there are two unknowns and many equations (depend on chosen window size). \( I_x(p_n) \) and \( I_y(p_n) \) are spatial derivatives and \( I_t(p_n) \) is time derivative of each pixel \( p_n \).

\[
\begin{bmatrix}
I_x(p_1) & I_y(p_1) & \cdots & I_t(p_1) \\
I_x(p_2) & I_y(p_2) & \cdots & I_t(p_2) \\
\vdots & \vdots & \ddots & \vdots \\
I_x(p_n) & I_y(p_n) & \cdots & I_t(p_n)
\end{bmatrix}
\begin{bmatrix}
u_x \\
u_y
\end{bmatrix}
= \begin{bmatrix}
I_t(p_1) \\
I_t(p_2) \\
\vdots \\
I_t(p_n)
\end{bmatrix} \tag{6}
\]

There are many approaches to solve system (6) but the most widespread are Lukas-Kanade [17] and Horn-Schunk [18] methods of optical flow estimation. Both methods assess displacement magnitude \( \vec{u} \) which minimizes error (7) for Lukas-Kanade and (8) for Horn-Schunk algorithm.

\[
E(u_x,u_y) = \sum_{i,j} g(x_i,y_j) \times \left[ u_x \frac{\partial I(x,y,t)}{\partial x} + u_y \frac{\partial I(x,y,t)}{\partial y} + \frac{\partial I(x,y,t)}{\partial t} \right]^2 \tag{7}
\]

where \( g(x_i,y_j) \) – two-dimensional Gaussian function.
\[
E(u_x, u_y) = \iint \left[ u_x \frac{\partial l(x,y,t)}{\partial x} + u_y \frac{\partial l(x,y,t)}{\partial y} + \frac{\partial l(x,y,t)}{\partial t} \right]^2 dx dy +
+ \lambda \iint \left( \frac{\partial u_x}{\partial x} \right)^2 + \left( \frac{\partial u_x}{\partial y} \right)^2 + \left( \frac{\partial u_y}{\partial y} \right)^2 + \left( \frac{\partial u_y}{\partial x} \right)^2 \right] dx dy
\]

Nakajima et al. [21], [22] used optical flow technique for estimation of human respiration parameters and posture changes. Authors proposed \(v^2\) parameter for evaluation of movement using video sequence. Proposed system detects posture change and respiration of the subject in bed by observing chest or blanket movement.

Kuo et al. [16] proposed a visual sleeping-respiration estimation system for monitoring and measuring the respiration parameters of sleeping people. Proposed system was built using near-infrared camera with a NIR projector. In their paper, authors evaluate new technique that combine Horn-Schunck optical flow estimation method and finite-state controlled hidden Markov model. The developed system can distinguish respiratory and non-respiratory body movement.

### 3.2. Respiration estimation by frame subtraction method

Frame subtraction that is usually applied for background removal [19], [20] can be used for respiration parameters estimation. The main feature of this method is removing regions that do not change from scene to scene. This technique allows deleting invariable background and leaving part of frame that has changed. Frame without background is computed using formula (9)

\[
l(x,y,t) = l(x,y,t) - l(x,y,t + 1)
\]

As seen from formula (9) the resulting frame \(\Delta l(x,y,t)\) contains region that was changed, on the black background (invariable pixel are canceled). Example of two frames subtraction is shown in Fig. 14.

**Fig. 13. Optical flow estimation for a subject in a supine position** [22]

**Fig. 14. Image subtraction:** (a) and (b) are the original images used for subtraction; (c) the result of subtracting (b) from (a) [23].
Tan et al. [23] presents system for respiration monitoring based on frame subtraction technique. Authors developed system for assessment breathing parameters. They made experiments to study dependence of respiration monitoring quality on the distance between object and camera. Dependence of closing contrast and the monitoring accuracy was researched.

Weixing et al. [24] and Ji et al. [25] proposed system for animal respiration monitoring based on frame subtraction technique. The proposed techniques can be used for identifying animal health in real-time by detecting breathing parameters.

3.3. Artefact removal in 2-D respiration extraction techniques

Despite the availability of 3-D and 2-D systems for non-contact respiration monitoring, they are far from wide commercial use. In particular, these systems are prone to artifacts of various origin, such as patient and background movement, ambient light change etc. Thus they cannot be used in clinical conditions without the guidance of qualified personnel. For accurate respiration monitoring of patient in supine position the facial artifacts removal is important as well. Mimics and eyes movement are the usual source of artifacts that should be canceled, since they introduce substantial distortion in scene leading to decrease of quality of respiration parameters extraction.

Systems for non-contact respiration monitoring using single optical sensor include one camera with required wavelength range, hardware for collecting and processing raw sensor data and software for video processing. Software implements estimation of movement in video sequence (hence possibility to compute respiratory parameters). Typical block diagram of respiratory monitoring system is shown in Fig. 15

![Fig. 15. Typical respiration monitoring system](image)

In this paper, facial artifacts removing is proposed to improve existent systems. Separate additional block is included in the block diagram. This block implements face recognition and exclusion of the face region from image to be analyzed at next stage. Algorithm is based on optical flow technique for motion estimation and Viola–Jones algorithm [26] for face region localization, and following removal of this area from video frame. The proposed method allows to remove artifacts induced by mimics movement and eye blinking. Modified block diagram of proposed system is shown in Fig. 16.

![Fig. 16. Respiration monitoring system with facial artifacts removal](image)

Procedure of data processing have the following stages: video flow capturing, motion estimation using optical flow technique, face recognition and removal of facial region from analysis, obtaining of respiration curve, target parameter calculation.

Conclusion

In this paper, the classification of respiration monitoring methods using optical sensors was proposed. Structure of respiration monitoring system with facial artifacts removing is developed, using the facial mimics and eyes blinking artefacts removal. New system can improve respiration monitoring for human positioned in frontal plane.

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Безконтактний моніторинг дихання з використанням оптичних сенсорів

Ціплю даної роботи є класифікація підходів до безконтактного моніторингу дихання і розробка структури системи моніторингу з усуненням артефактів міміки. Усі наведені методи були розділені на дві основні групи: методи на основі визначення дихання з 3-D зображення об’єкта і методи на основі 2-D обробки зображень. Була розроблена структура системи моніторингу дихання на основі оптичних сенсорів з можливістю видалення артефактів міміки. Новий підхід дозволяє покращити моніторинг дихання для об’єктів в положенні лежачи на спині і в позиції сидячі.

Ключові слова: безконтактні вимірювання, безконтактний моніторинг дихання, моніторинг дихання, артефакти міміки, оптичний потік.

Бесконтактный мониторинг дыхания с использованием оптических датчиков

Целью работы является классификация подходов к бесконтактному мониторингу дыхания и разработка структуры системы мониторинга с устранением артефактов мимики. Все имеющиеся методы были разделены на две основные группы: методы на основе определения дыхания из 3-D изображения объекта и методы на основе 2-D обработки изображений. Была разработана структура системы мониторинга дыхания на основе оптических датчиков с возможностью удаления артефактов мимики. Новый подход позволяет улучшить мониторинг дыхания для объектов в положении лежа на спине и в позиции сидя.

Ключевые слова: бесконтактные измерения, бесконтактный мониторинг дыхания, мониторинг дыхания, артефакты мимики, оптический поток.

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