Endoscopic Navigation for Minimally Invasive Suturing

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Abstract. Manipulating small objects such as needles, screws or plates inside the human body during minimally invasive surgery can be very difficult for less experienced surgeons, due to the loss of 3D depth perception. This paper presents an approach for tracking a suturing needle using a standard endoscope. The resulting pose information of the needle is then used to generate artificial 3D cues on the 2D screen to optimally support surgeons during tissue suturing. Additionally, if an external tracking device is provided to report the endoscope's position, the suturing needle can be tracked in a hybrid fashion with sub-millimeter accuracy. Finally, a visual navigation aid can be incorporated, if a 3D surface is intraoperatively reconstructed from video or registered from preoperative imaging.

1 Introduction

Surgical interventions are increasingly being performed in a minimally invasive fashion. The main advantages of causing less trauma to the patient, smaller infection rates, and faster recovery are well documented. Minimally invasive surgeries (MIS), however, require a lot of experience and pronounced skills from the surgeon. One of the main reasons lies in the almost complete loss of depth perception that makes the manipulation of small objects such as needles, screws, or plates very difficult. This paper presents a framework for tracking small objects and estimating their pose using a standard, monocular endoscope. As one example application, the tracking and pose estimation of a suturing needle is presented. The resulting pose information is then used for augmenting the surgeon's view by generating artificial 3D cues onto the 2D display and hence providing additional support during suturing. Such augmentation techniques are especially helpful for trainees and less experienced surgeons. The presented approach is general and can be applied to any object with known geometry by performing minor adaptations of the pose estimation module.

Traditional setups usually cannot support the navigation process, as it is often impossible to track small objects due to the missing line of sight and/or because their dimensions or function does not allow marker attachment. While magnetic markers of small dimensions exist, the presence of metallic objects

in their vicinity severely deteriorates their accuracy. Our method can offer a solution by relying on an external tracking device reporting the endoscope's pose. Based on this information the needle can be fully tracked in a hybrid fashion in the world coordinate system. Finally, if pre- or intra-operative data are registered to the system, the possible interaction between the tracked object and the registered data can also be visualized and used as a navigation aid.

The methods proposed in this paper rely on the combination of object tracking, pose estimation and view augmentation. Color-coded instrument tracking has been presented in [1] for positioning surgical instruments in a robotized laparoscopic surgical environment. Another color coded approach for tracking laparoscopic instruments was proposed in [2]. The pose computation from objects in 2D images given 2D to 3D point correspondences has been explored in [3], using textures in [4] and from parameterized geometry in [5]. Pose computation from different geometric entities such as ellipses, lines, points and their combinations was presented in [6].

An endoscopic hybrid tracking approach used for spine surgery has been presented in [7]. Passive fiducials are attached to the vertebrae and their pose detected by a tracked endoscope. However, the registration of the fiducials with the target anatomy adds another step to the navigation procedure. In [8], fiducials are attached to the tools and their 3D pose measured by a head-mounted monocular camera. The results are displayed using a HMD. Objects without a direct line of sight to the camera cannot be handled using this system. Other approaches for tracking and pose estimation of surgical instruments include structured light endoscopes [9] and ultrasound probes [10]. An approach for facilitating the depth perception was presented in [11], where the invisible shadows of the tools were artificially rendered into the scene.

We propose a system for estimating the pose of very small colored objects in order to augment the surgeon's view by relying only on the standard environment of endoscopic surgeries. No modifications of the instruments involved (like marker attachment) is needed, only the surface color of the object to be identified has to be adjusted. This approach enables tracking of these objects in a hybrid fashion and enhances existing navigation systems.

2 Methods

For all experiments a 10 mm radial distortion corrected endoscope¹ with an oblique viewing angle of 25° was used. In order to avoid interlacing artifacts, a progressive frame, color CCD camera with a resolution of 800 × 600 pixels and 30 fps has been incorporated. The camera can be calibrated pre-operatively by the surgeon without requiring technical assistance [12]. As the endoscope/camera combination provides a depth of field in the range of 3 – 7 cm, the focal length of the camera can be kept constant during the entire procedure. This allows to avoid recalibration of the system during surgery.

¹ Panoview Plus, Richard Wolf GmbH, <http://www.richard-wolf.com/>

The surgical needle² has a nearly half circular shape with a radius $r = 8.5$ mm and a needle diameter d of 1 mm, see Fig. 4b. Tracking such a needle in a surgical environment is a challenging task due to the moving camera, the cluttered background, and the unconstrained motion. In addition, partial occlusions of the needle occur during the suturing process and while being held by the gripper. Finally, the coaxial illumination yields strong specularities and an inhomogeneous light distribution. The needle was painted using a light matt green color finish, allowing robustly dealing with the problems listed above.

On startup, the system enters an interactive phase, where the needle first needs to pass through the central search area in the image. Once the needle is found, the tracking and pose estimation starts. Due to the inherently ambiguous solution of the orientation of the needle, the surgeon quickly needs to verify whether the correct orientation has been chosen and manually switch to the correct solution if necessary. The whole algorithm is depicted in Fig. 1 and presented in more detail in the following paragraph.

Fig. 1. Needle detection algorithm after interactive initialization

A color segmentation is applied to the search area, selecting all green pixels. The *RGB* image is converted to the *HSI*-color space and all pixels within the hue range from $65^\circ \leq H \leq 160^\circ$ are selected. Additional constraints for the saturation $S \geq 0.15$ and intensity $I \geq 0.06$ reduce the specularities and help to remove dark areas. All of these values were empirically determined but remained constant through all experiments. The segmented pixels are labeled using a connected component labeling algorithm while small components of less than 50 pixels are discarded. An ellipse is fitted to all remaining components using the method proposed in [13]. This results in the center of the ellipse $\mathbf{c}_e = [x_0, y_0]^T$, the major and minor axis a and b , and the tilt angle θ , defined as the angle between the major axis a and the x -axis, as shown in Fig. 2a. The actual parameter set $\mathbf{p}_t = [x_0, y_0, a, b, \theta]^T$ is compared to the previously computed set $\mathbf{p}_{t-1}$. If the new ellipse center is within the bounds defined by the predicted motion and the change of the parameters a , b and θ is below 30 %, the new model is regarded as valid and a new motion vector is computed for adjusting the search area in the next image. Otherwise the found ellipse is discarded, the next image acquired, and the ellipse's parameters are estimated under the same conditions

² Atraloc, Ethicon Ltd., <http://www.ethicon.com>

Fig. 2. a) 2D Ellipse parameters $c_e = [x_0, y_0]^T$, a , b , θ ; b) projection of a 3D circle, defined by $C = [X_c, Y_c, Z_c]^T$, n , to the image plane

as before. From the found 2D ellipse parameters, the corresponding 3D circle is computed using the method from [14]. This results in the location of the circle center $C = [X_c, Y_c, Z_c]^T$ and the normal n of the plane containing the needle as depicted in Fig. 2b. The computation of the needle pose is, however, inherently ambiguous and results in two distinct solutions for the center (C' , C'') and for the normal (n' , n''). The correct circle center can be determined by backprojecting both solutions $c = PC$ to the image and choosing the solution yielding the smaller distance to the previously computed ellipse center c_e , with P being the projection matrix of the camera containing the intrinsic parameters. In order to choose the correct circle plane n_t , the current plane normals (n' , n'') are compared to the previously computed normal n_{t-1} and the solution with the smaller angular difference $d = \cos^{-1}(n_t \cdot n_{t-1})$ is used. In cases where the needle passes through a degenerate pose as shown in Fig. 3d, the current normal is selected to move consistently with the prior motion.

The methods presented above can be used to compensate the loss of 3D depth perception by rendering artificial orientation cues. This does not require significant adaptations from the surgeon as his working environment remains unchanged. For example, a semi-transparent plane containing the needle can be projected onto the 2D image indicating the relative pose of the needle with respect to the camera as can be seen in Fig. 3. The plane is represented by the square surrounding the circle containing the needle, whose vertices X_i are backprojected to the image $x_i = PX_i$. The square is displayed in a semi-transparent way in order to minimize the loss of information due to occlusion and its projective distortion indicates which part of the needle is closer to the camera.

As mentioned above, the system described can be extended to allow object tracking, too. If the endoscope is tracked externally and the needle pose is computed in the camera coordinate system, as indicated in Fig. 4a, hybrid tracking of the needle is possible: $H_{world}^{needle} = H_{cam}^{needle} H_{marker}^{cam} H_{world}^{marker}$, with H_{world}^{marker} the pose of the marker in the world coordinate system, H_{marker}^{cam} the transformation relating the camera with the world coordinate system as computed during the calibration process, and H_{cam}^{needle} the resulting transformation from the pose estimation process as described above.

For these experiments, a marker is attached to the endoscope that is being followed by the active optical tracker³ providing accurate position (< 0.2 mm

³ EasyTrack500, Atracsys LLC., <http://www.atracsys.com>

Fig. 3. a) Example visualization of the needle showing the detected ellipse and the plane, b) needle held by gripper, c) example showing partial occlusion during the suturing process on a phantom mockup, d) degenerate case with the needle plane being almost perpendicular to the image plane

localization error) and orientation information in a working volume of $50 \times 50 \times 50 \text{ cm}^3$. An external hardware triggering logic ensures the synchronized acquisition of the tracking data and the camera images during dynamic freehand manipulation.

Fig. 4. a) Overview of the different components of the presented system, b) close-up of the needle held by a gripper

Even more navigation cues can be integrated if registered 3D data such as CT or MRI are available and if the needle is tracked. This allows to visualize the possible interactions between the tracked object and the 3D data resulting in a navigation aid. In this paper, such static 3D models are created by the same endoscope, using an intra-operative 3D reconstruction method, as described in [15]. As the endoscope is moved over the surgical scene, the images from this sequence are used to build a 3D model of the visible anatomy online during the intervention, resulting in a 3D triangle mesh M_{world} . The plane π_{world} defined by (C, n) in the world coordinate system can then be cut with the 3D triangle mesh M_{world} resulting in a lineset $l = \pi_{world} \cap M_{world}$. A visibility filter v_{cam} returns those lines from this set that are visible in the current view: $l_{vis} = v_{cam}(l, M_{world}, [R, t])$, with $[R, t]$ being the camera pose in the world coordinate system. v_{cam} casts rays from the camera position to both vertices of each line. If a face is encountered between the camera position and one of the line vertices, this line is set to invisible. This pessimistic approach may lead to the complete

Fig. 5. a) 2D augmented view with needle plane and cutting line, b) internal 3D representation of the same scene showing the intra-operatively reconstructed texture-mapped 3D model, a disk representing the needle and the 3D cutting line

loss of only partially visible line segments. However, this is an improbable event considering usual organ topography. In rare cases where multiple cutting lines result from this procedure, the solution closer to the needle is selected. Please note that, while the reconstructed 3D model is essential for these operations, it remains hidden from the surgeon and the system only presents the final result (i.e. the cutting line), overlaid on the endoscopic view (see Fig. 5a).

3 Results

The framerate for the hybrid tracking is mainly determined by the needle detection and the ellipse fitting process. They both depend on the number of pixels thus the processing time decreases with the distance between the needle and the camera. In the working range of 3 – 7 cm the system runs in real-time and the framerate on a 2.8 GHz CPU is between 15 – 30 fps. The framerate for the virtual interaction depends on the number of vertices and faces of the 3D model. For the 3D models in our experiments with approximately 500 – 1000 faces the framerate dropped to 10 – 15 fps.

In order to assess the accuracy of the pose estimation, a 2 DOF positioning table with a resolution of 0.02 mm was used to move the needle to distinct positions, covering the whole working volume, as depicted in Fig. 6a. As the errors in x and y are very similar (both depend mainly on the in-plane resolution of the camera), only the errors in x and z are presented here. Around each distinct position, defining the reference coordinate system eight shifts of 0.1 mm were introduced and the pose estimated by the system. The measurements were then compared to the manually set (ground truth) position. The standard deviations of the errors in the $z = 30$ mm plane were 0.003 ± 0.065 mm along the optical z -axis and -0.01 ± 0.018 mm along the x -axis. On the lateral positions the errors were -0.02 ± 0.024 mm along the optical z -axis and -0.01 ± 0.015 mm along

Fig. 6. a) Setup for measuring the errors, b) needle plane parallel to image plane and with an inclination of 60°

the x -axis. For the $z = 70$ mm plane, the errors along the optical axis were 0.053 ± 0.05 mm in the z -direction and 0.04 ± 0.03 mm in the x -direction. The measurements on the lateral positions resulted in errors of 0.037 ± 0.024 mm and 0.026 ± 0.043 mm respectively.

The angular accuracy was quantified in a similar way as in the above experiment by introducing controlled rotations (5°) of the needle with respect to the image plane. This has been repeated with the needle being parallel to the image plane, and inclined by 30° and 60° respectively. For the needle plane being parallel to the image plane, the standard deviations for the angular errors were $2.3 \pm 0.5^\circ$ in the $z = 30$ mm plane and $2.5 \pm 0.5^\circ$ in the $z = 70$ mm plane. For the needle plane inclined by 30° the standard deviations were $1.1 \pm 0.7^\circ$ close to the camera and $1.3 \pm 0.7^\circ$ at the farthest position. Finally, for the 60° case, the errors were $0.7 \pm 0.5^\circ$ and $1.0 \pm 0.5^\circ$ respectively. As expected from the projective nature of the image formation, the error decreases with the increasing inclination of the needle plane.

4 Conclusions

In this paper a multi-purpose tracking system for small artificial objects was presented that can cope with partial occlusions, cluttered background and fast object movement. The proposed system allows real-time tracking of a suturing needle with sub-millimeter accuracy. The needle tracking is very robust and quickly recovers from occlusions, however their influence on the accuracy still needs to be carefully investigated. The augmented visualization helps the surgeon to perceive 3D cues from 2D images using the existing instrumentation. The hybrid tracking can improve existing navigation systems by adding the possibility to handle small objects.

The color dependence of the system requires the white balance to be set accurately in advance, which has to be integrated into the calibration procedure. The pose ambiguity cannot always be resolved correctly, therefore the system allows to manually switch to the correct solution. It is planned to solve this ambiguity by using other visual cues.

Future work includes tracking and pose estimation for other objects such as screws and plates leading to a greater flexibility of the system as well as enabling

to simultaneously track multiple objects. The most important step to come is the clinical validation of the system and its impact on the surgeons' performance. We are currently setting up studies for this purpose, which will also enable us to investigate, which augmentation technique offers the most benefits.

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