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2 Projection-based polygon estimation in X-ray computed 3 tomography 4 5

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9 **Abstract**

10 *In X-ray computed tomography, the 3D structure of a scanned object can be*
11 *reconstructed from a number of projection images of the object acquired from different*
12 *directions. Conventional tomographic reconstruction algorithms represent the*
13 *reconstructed volume on a voxel grid. Such representation is, however, not well suited*
14 *for polyhedral objects arising in many industrial applications, since such objects are*
15 *first voxelized during the reconstruction and then processed in order to obtain a polygon*
16 *mesh representing the surface of the object. These transformations lead to loss of details*
17 *and may induce artefacts that hinder posterior image processing.*

18 *In this work, a new approach is proposed in which a contour polygon of the object is*
19 *directly estimated from the projection data. The approach is based on simulated*
20 *projections of the polygon model and optimization of the vertex positions in the model*
21 *with respect to the distance between the simulated and the original projection data*
22 *(projection distance). The obtained results demonstrate the ability of the proposed*
23 *algorithm to accurately represent the contour of the object even in case of noisy*
24 *projection data.*
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41 **Introduction**

42 X-ray computed tomography is an imaging technique that is capable of revealing the 3D structure
 43 of a scanned object with numerous applications in medicine, research (e. g., materials research),
 44 and industry (e. g., quality control). A typical workflow in industrial quality control involving
 45 computed tomography includes manufacturing of an object according to a CAD model, the
 46 reconstruction of the object using a tomographic reconstruction algorithm, segmentation,
 47 estimation of a polygon mesh representing the surface of the segmented object, and comparison
 48 with the CAD model used during manufacturing [1,2].

49
 50 Conventional tomographic reconstruction algorithms, such as FBP [3], SIRT [4], or DART [5],
 51 represent the reconstructed volume on a voxel grid. This representation is not well suited for
 52 polyhedral objects arising in many industrial applications, since such representation can lead to
 53 loss of details in the object and to introduction of image artefacts [6]. Moreover, artefacts in the
 54 reconstructed volume can hinder posterior processing [1].

55
 56 To generate a polygon mesh of the surface, contouring techniques, such as Marching Cubes [7]
 57 and its modifications, are widely used. These techniques create a triangular mesh using
 58 interpolation between attenuation values calculated on the voxel grid. Recently, a technique [8]
 59 has been proposed, where reconstruction on an iteratively deformed tetrahedral mesh is used
 60 instead of reconstruction on a regular voxel grid. These methods, however, are sensitive to noise
 61 and reconstruction artefacts and require a high resolution reconstruction to reproduce sharp edges
 62 on an object [7,8].

63
 64 In this work, a new approach is proposed in which parameters of a model representing the object
 65 are directly estimated from the projection data, eliminating need for reconstruction on the pixel
 66 (voxel) grid and therefore decreasing related drawbacks. Our approach is based on optimization
 67 of an analytical model of the object with respect to a difference between the simulated projections
 68 of the model and the original projection data. Such models are readily available in many
 69 application domains, e. g., in industrial quality control, where objects being controlled are
 70 manufactured according to CAD models. In the present paper, we focus on the two-dimensional
 71 case and use polygons as the models to estimate the contour of the object. Two particular
 72 problems, *Edge estimation* and *Polygon estimation*, relevant for diamond processing, are
 73 considered.

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76 **Method**

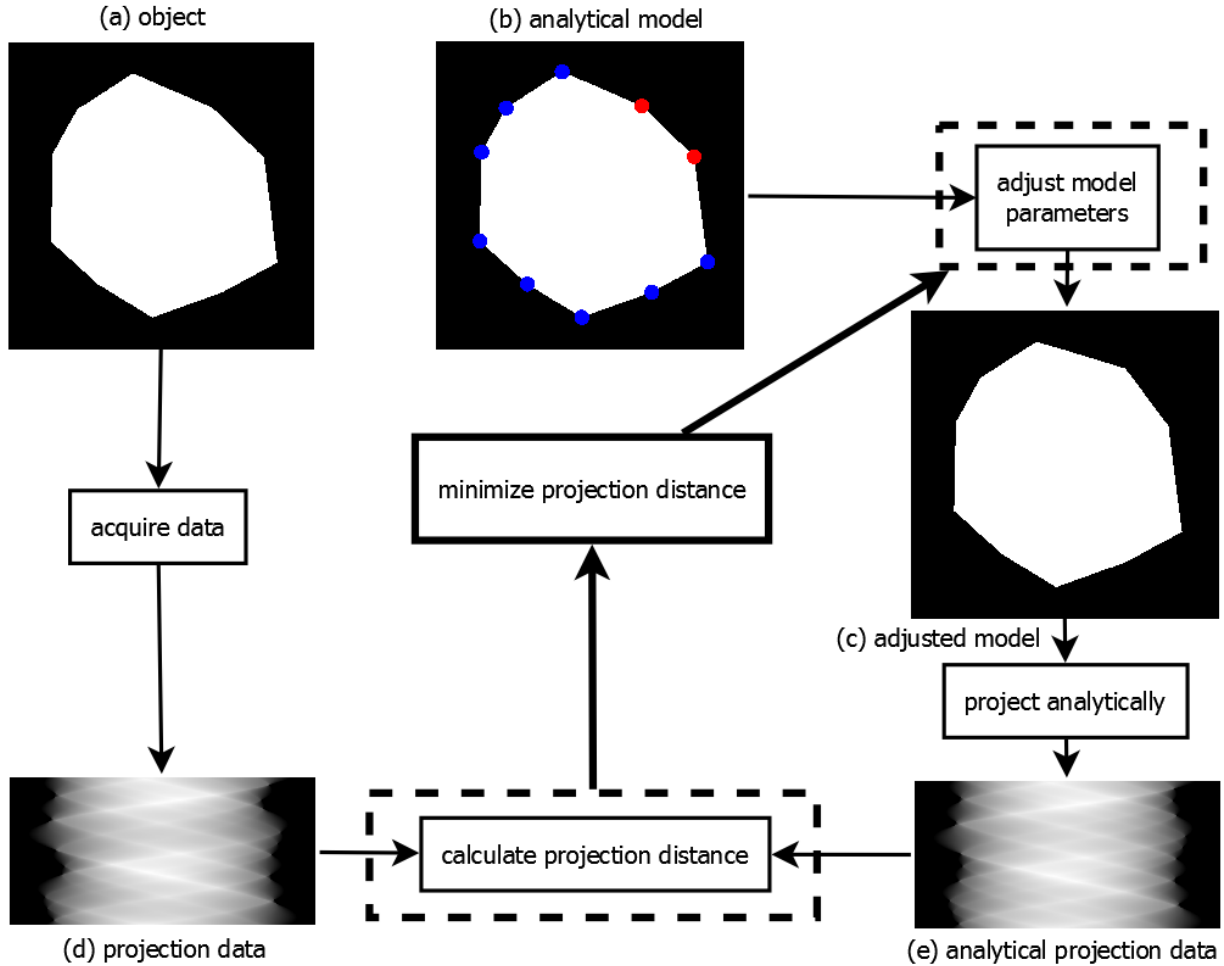
77 In tomography, the projection process can be modeled as a linear operator determined by the
 78 projection geometry, which leads to a system of linear equations

$$79 \quad Wx = p, \quad (1.1)$$

80 where $p \in R^m$ is the projection data, $x \in R^n$ is an image representing the object on a pixel
 81 (voxel) grid, $W \in R^{m \times n}$ is the projection matrix with m being the number of detector elements
 82 multiplied by the number of projection angels and n being the number of pixels in the image. For
 83 any image $s \in R^n$ we can define the difference between its simulated projections and the
 84 measured projection data as

85
$$d(s) = \|Ws - p\|_2, \quad (1.2)$$

 86 which is known as the *projection distance*. Analogously, a projection operator W_M for a model-
 87 based representation of the object and the projection distance $d_M(t) = \|W_M t - p\|_2$ for any model
 88 t can be defined.
 89

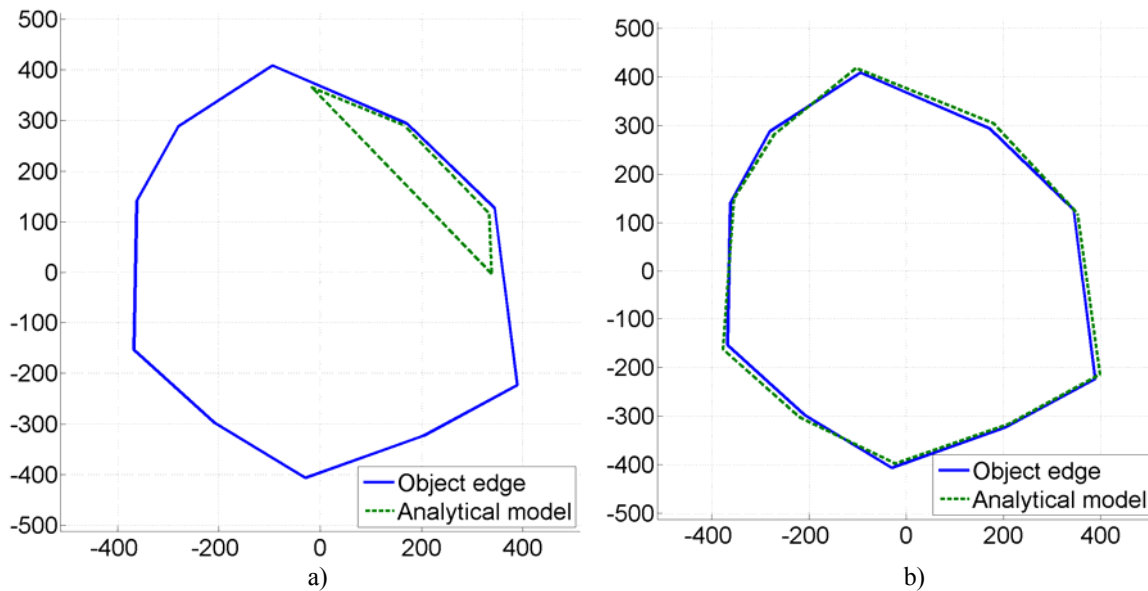


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 91
 92 **Figure 1:** Schematic overview of the proposed model estimation approach.
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95 Our approach is to find the parameters of the model t of the object that minimize the projection
 96 distance $d_M(t)$ between the acquired projection data and simulated projections of that model.
 97

98 Consider projection data p (Figure 1d) which was acquired from the unknown object
 99 (Figure 1a). The approach consists in adjusting parameters of a model of the object (Figure 1b),
 100 which is then analytically projected. Next, the obtained projection data of the model (Figure 1e)
 101 is compared to the measured projection data of the object using the projection distance. The
 102 parameters that minimize the projection distance are retained as the parameters of the model
 103 representing the object.

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 105 The above described approach is very general and can be applied to a variety of models
 106 representing objects arising in different domains where computed tomography is applied. In the
 107 present paper, we consider two particular problems relevant for diamond processing.
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 112 **Figure 2:** (a) *Edge estimation* and (b) *Polygon estimation*.
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115 In *Edge estimation* (Figure 2a), we aim at estimating the slope of one edge of the object assuming
 116 that an approximate position and slope of that edge is available and there is no prior model of the
 117 object. A trapezium is used as an analytical model, parameterized with rotation, shift, length of
 118 the top basis and the base angles. The optimization of the projection distance is performed only
 119 for a part of the available projection data that is located in proximity to the projections of the
 120 edge in question and corresponds to X-rays being roughly parallel to it.
 121

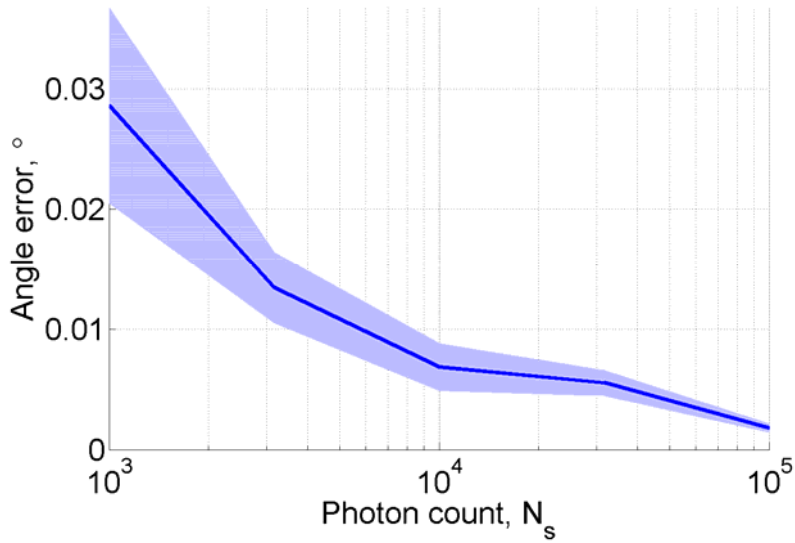
122 In *Polygon estimation* (Figure 2b), a complete object is estimated assuming the availability of a
 123 prior polygonal model representing the object. The optimization is iterative, and during each
 124 iteration each vertex is adjusted while keeping the other vertices fixed. This procedure allows to
 125 split the optimization in a high-dimensional search space into several optimizations in two-
 126 dimensional search spaces.
 127

128 In both cases, we assume that the objects are homogeneous without any holes and that their
 129 attenuations are known. To optimize the projection distance, an interior-point method for
 130 nonlinear programming [9] is used. It is an iterative algorithm that combines a line search method
 131 computing steps by factoring the primal-dual equations and a trust region method employing
 132 conjugate gradient iterations. Bounds for the parameters of the model can be used to restrict the
 133 search space. These bounds depend, in principle, on the accuracy of the prior model.
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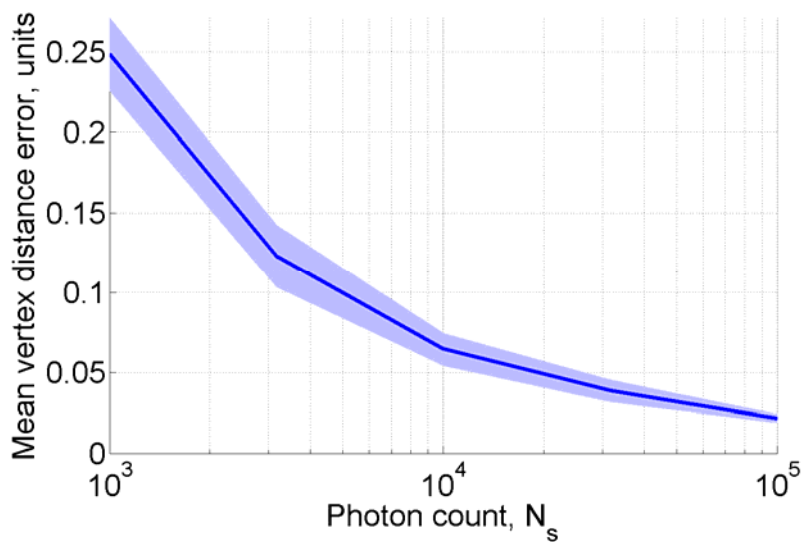
136 Experiments

137 *Simulation experiments*

138 A number of experiments were set up by simulating projection data of a polygonal phantom to
 139 demonstrate the proposed approach. In all experiments described in this section, a detector with
 140 1044 elements was used, 500 projections were computed and Poisson noise was added to obtain
 141 5 datasets for each noise level.
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 145 **Figure 3:** Angle error as a function of the photon count for the edge estimation.
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 150 **Figure 4:** Mean vertex shift error as a function of the photon count for the polygon estimation.
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153 In the first series of experiments, the ability of the proposed approach to handle Edge estimation
154 problem was evaluated. For each noisy dataset computed for the object shown in Figure 2a, the
155 slope of one edge was estimated as described in Section Method, and the mean angle error was
156 plotted (Figure 3) together with the standard error (shown as shaded area in the plot).

157
158 In the second series of experiments, the polygon representing the object was estimated using the
159 prior model (shown in Figure 2b together with the object), which was obtained from the object by
160 randomly shifting each vertex, the mean shift was 12.17 units. For each noisy dataset the vertex
161 positions were estimated, and the mean distances between the true and estimated vertex positions
162 were calculated (Figure 4).

163
164 The results demonstrate the ability of the proposed approach to achieve subpixel accuracy in
165 polygon estimation (Figure 4) and to accurately determine edge slopes (Figure 3), confirming that
166 the proposed approach can accurately estimate parameters of a model representing the object
167 based on the projection data even in the presence of noise.

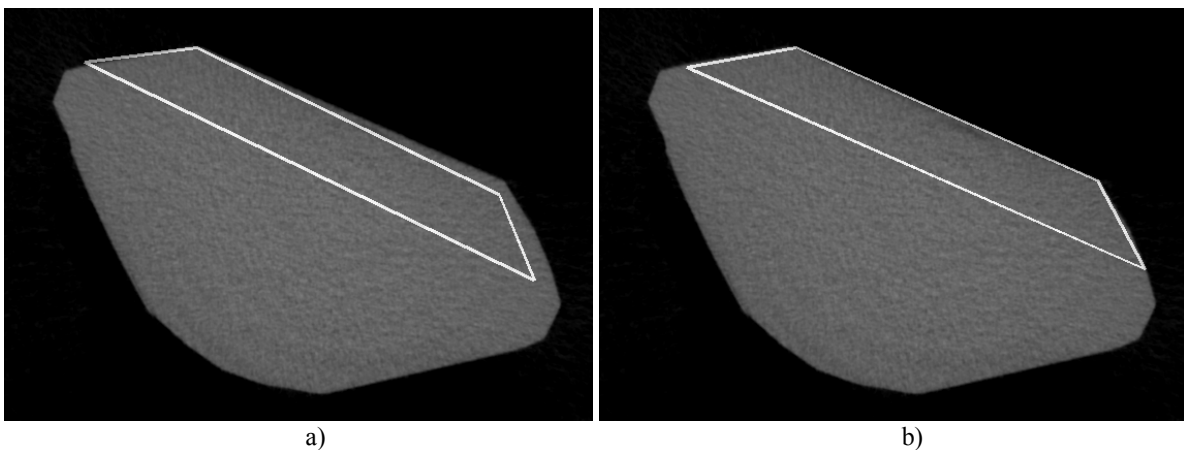
168

169 *Real experiments*

170 Two cone-beam datasets were acquired using a desktop micro-CT system SkyScan-1172
171 (Bruker-MicroCT, Belgium), size of each projection image was 1000×524. The objects were
172 known to be homogeneous, their attenuation was calculated as the mean attenuation in an inner
173 region. One slice from each dataset was used as the projection data to demonstrate the
174 performance of the proposed approach on real data.

175
176 For the first experiment, 280 projections of a diamond with a polished facet were acquired. The
177 proposed approach was applied to estimate the slope of the polished facet, for which an initial
178 model was roughly determined using SIRT reconstruction of the same slice (Figure 5a). The
179 obtained result is shown in Figure 5b.

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Figure 5: Facet estimation in a diamond: SIRT reconstruction overlaid with initial (a) and final (b) models.

187 In the second experiment, 515 projections of an almost polished diamond were acquired. The
 188 polygon representing an initial model was again roughly determined using SIRT reconstruction
 189 (Figure 6a) and supplied into the proposed approach. The resulting polygon is presented in
 190 Figure 6b. While the final model may appear to slightly overestimate the object, e. g., in the
 191 bottom right corner of the figure, enlarged plots (Figures 6c, 6d) suggest the presence of material
 192 there, and the lower attenuation of the object in that area might have been caused by a
 193 reconstruction artefact.
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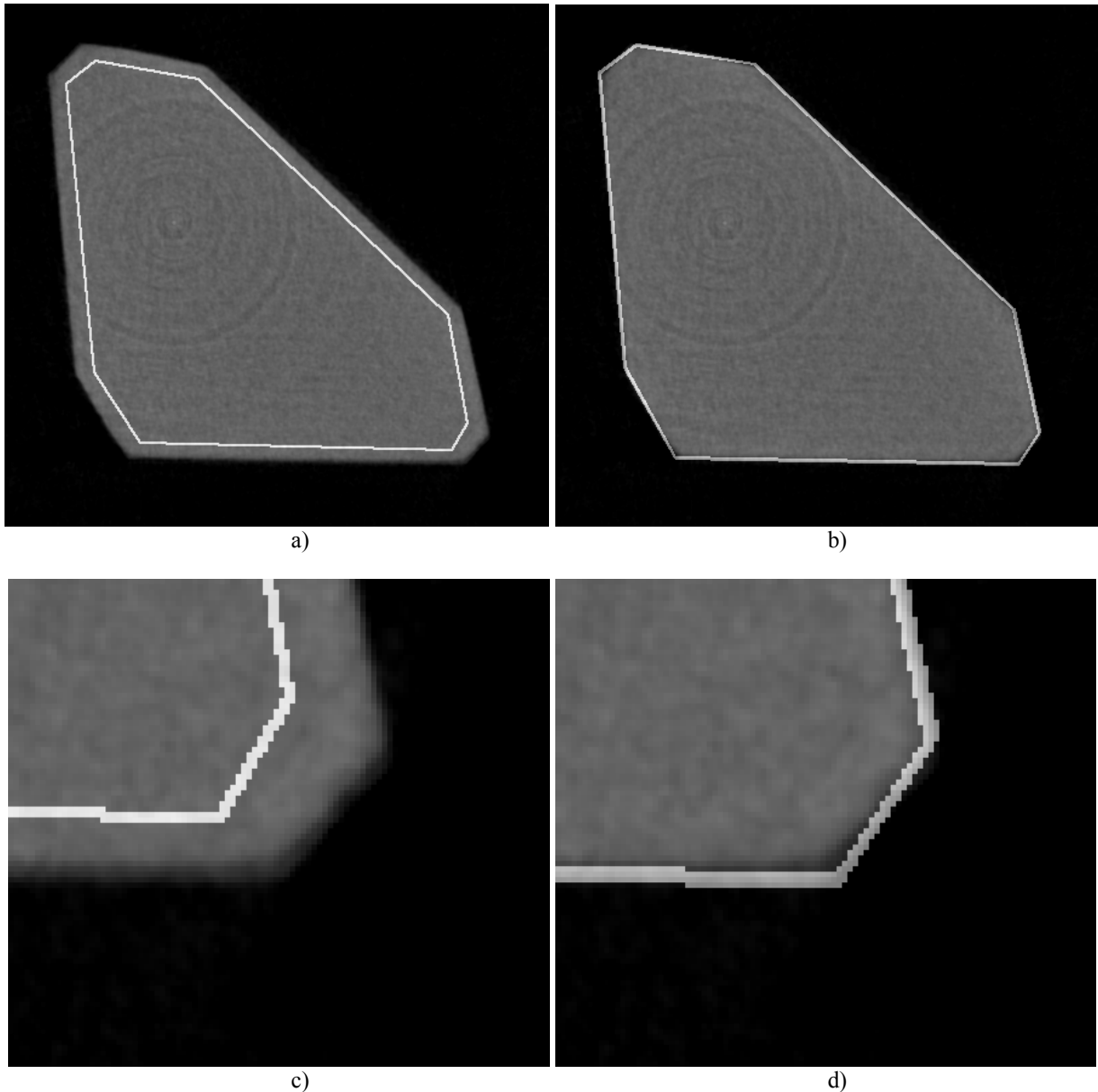


Figure 6: Polygon estimation in a diamond: SIRT reconstruction overlaid with initial (a) and final (b) models and the corresponding enlarged fragments (c) and (d).

205 The proposed model estimation approach demonstrates plausible and promising results on the
206 real datasets, allowing to compute the parameters of the models representing the scanned objects,
207 and is potentially robust against reconstruction artefacts. Nevertheless, further validation of this
208 approach is required.

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211 **Discussion and conclusion**

212 In this paper, a projection-based polygon estimation approach for X-ray computed tomography
213 was proposed. In this approach, a model representing the contour of the object is directly
214 estimated from projection data, eliminating need for reconstruction on the pixel (voxel) grid.
215 While polygon models were used to represent the objects throughout the paper, this technique is
216 readily extensible to other representations, such as splines.

217
218 Experiments show that the proposed approach can accurately estimate the parameters of the
219 models representing the objects even in the presence of noise and has great potential to enhance
220 diamond processing or quality assessment of industrial parts.

221
222 Future work will focus on the extension of the proposed approach to three-dimensional models
223 and non-homogeneous objects and further validation on real datasets.

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226 **Acknowledgements**

227 The authors would like to thank Diamcad (Belgium) for providing the diamond datasets. This
228 work was financially supported by the BOF LP project 25778 and the iMinds-MetroCT project
229 (iMinds, Interdisciplinary Institute for Technology, is a research institute founded by the Flemish
230 Government).

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