

Effortless Scanning of 3D Object Models by Boundary Aligning and Stitching

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Abstract: We contribute a novel algorithm for the digitation of complete 3D object models that requires little preparation effort from the user. Notably, the presented algorithm, Joint Alignment and Stitching of Non-Overlapping Meshes (JASNOM), completes 3D object models by aligning and stitching two 3D meshes by the boundaries and does not require any previous registration between them. JASNOM only requirement is the lack of overlap between meshes, which is simple to achieve in most man made object. JASNOM takes advantage that both meshes can only be connected by their boundary to reframe the alignment problem as a search of the best assignment between boundary vertices. To make the problem tractable, JASNOM reduces the search space considerably by imposing strong constraints on valid assignments that transform the original combinatorial problem into a discrete linear problem. By not requiring previous camera registration and by not depending on shape features, JASNOM contributions range from quick modeling of 3D objects to hole filling in meshes.

1 INTRODUCTION

We propose an algorithm, Joint Alignment and Stitching of Non-Overlapping Meshes (JASNOM), that requires little preparation and technical knowledge to create a 3D model. JASNOM exploits the underlying manifold structure of range sensors data to recreate the object surface from just two range images.

Obtaining a pair of meshes that comply with these constraints can be easily achieved using active 3D cameras such as the Kinect camera. Since mesh boundaries are typically in regions of strong curvature, e.g., corners and edges, they do not change considerably under small perturbations on the view point. Thus non-overlapping meshes can be obtained by simply flipping objects, as illustrated in Figure 1, or roughly positioning two cameras in opposite directions of the object for non-rigid objects.

By not requiring a-priori camera registration nor extra apparatus, JASNOM provides a simplified process for object modeling. Furthermore, by using the boundary geometry for aligning meshes, JASNOM does not depend on geometric nor texture feature matching. In this work we illustrate the potential for fast object modeling using a non rigid object, a Human, and different small hand made objects.

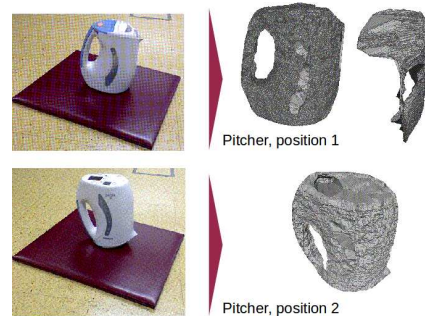


Figure 1: Example of a possible, and effortless, procedure for acquisition of two non-overlapping meshes using a Kinect sensor.

Another possible application of JASNOM is to fill holes in a mesh. In the case of interactive object modeling, our algorithm allows a user to select parts from a mesh or library of meshes and use them to fill holes in an incomplete 3D model. The possibility of filling holes from other mesh parts is of valuable use for modeling objects with self similar surfaces such as planes or cylinders, which are the basic shapes of the man-made objects that populate indoor environments.

JASNOM addresses jointly both the problem of registration and merging of meshes by aligning two

meshes by their boundary. As depicted in Figure 2, JASNOM aligns two meshes, M_1 and M_2 , and glues them to create a single mesh, M . While JASNOM applications can be extended to any problem that can be formulated by boundary alignment, e.g., puzzles, JASNOM was developed with a primary focus on 3D object modeling.

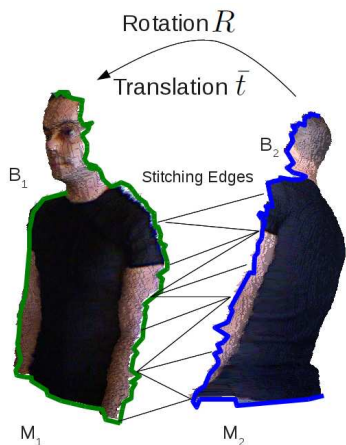


Figure 2: Our objective is to construct a mesh M from two other meshes, M_1 and M_2 . Both meshes have a boundary B_1 and B_2 that do not overlap. To construct M , we align both boundaries through a rotation R and a translation \bar{t} .

JASNOM aligns meshes by assuming that their boundaries are the same geometric structure but seen in different reference frames, i.e., that each point in one boundary has a corresponding point in the other. Under this assumption, stitching edges should connect corresponding vertices in the two boundaries and should have zero length. The stitching problem can then be posed as that of finding correspondences between boundaries and the aligning problem as that of finding the rigid transformation that minimizes the total edge length.

However, in a realistic scenario, the boundaries do not exactly match and there is no a-priori knowledge on the correspondences between the boundaries. In this case, the previous solution would have three main drawbacks: i) if the boundaries are strongly irregular, simple minimization of edge lengths may lead to intersections between meshes; ii) in general, finding correspondences between vertices is a combinatorial problem; iii) there is no guarantee that the correspondences by themselves will define a triangular mesh that allows the completion of the mesh.

Our main contributions address these problems and allow the reconstruction of a triangular mesh between the two boundaries. Namely JASNOM:

- introduces a cost function that penalizes both the edge lengths and the intersection between meshes;

- introduces constraints that simplify the search for the assignments from a combinatorial problem to a discrete linear programming problem, solvable in linear time;
- introduces a stitching algorithm that reconstructs the triangular mesh given a set of assignments.

JASNOM penalizes the intersection between meshes by modeling the intersection as a set of local conditions to be verified by each stitching edge. Section 3 introduces the cost function and our minimization strategy.

To constrain the search space for the assignments, JASNOM uses the fact that the resulting mesh should have the same properties as an object surface. E.g., object surfaces are 2D-manifolds and thus object surface meshes cannot have edges crossing each other except at vertices. Section 4 addresses some properties of object surface meshes and their impact on establishing assignments between boundaries.

To reconstruct the mesh structure, JASNOM makes use of the assignments from the alignment stage and ensures that properties like mesh *manifoldness* are locally preserved. Section 5 addresses the problem of reconstructing meshes from assignments.

Section 6 illustrates different applications of JASNOM and Section 7 concludes.

2 Related Work

The use of range images for 3D object modeling motivates the use of mesh stitching to construct complete models. Due to their planar topology, range images induce an intrinsic mesh in point clouds, but do not represent the whole object. Thus, to recover the complete object surface, several meshes from range images can be stitched together, instead of using point cloud filtering approaches, such as (Kazhdan et al., 2006), (Levin, 2003), (Guennebaud and Gross, 2007) or (Newcombe et al., 2011), or using approximations to the convex-hull, such as (Edelsbrunner and Mücke, 1994) or ball pivoting (Bernardini et al., 1999). In terms of applications, we note that using the original mesh and vertices instead of using pos-processing approaches introduces several advantages, e.g., adding color to the models is immediate. As such, we here focus on other works that preserve the original mesh.

In (Turk and Levoy, 1994), authors present a three step algorithm for stitching range images that makes use of the overlap between images to both align and stitch them. The algorithm first step is to align meshes by means of an Iterative Closest Point (ICP) algorithm, (Besl and McKay, 1992). The second step

removes overlapping regions between two adjacent meshes, by deleting triangles. This step leaves only the triangles that do not overlap or that overlap only partially. The final step stitches meshes by the points where the partially overlapped triangles intersect. The stitching procedure adds vertices at the intersection and new triangles are built on top of the original ones. When there is no overlap, meshes cannot be aligned using ICP and the stitching cannot be built on top of existing triangles.

More recently, different authors, e.g. in (Marras et al., 2010) and (Pauly et al., 2005), used the technique described in (Turk and Levoy, 1994) for mesh stitching with the purpose of filling holes in a model. In both algorithms an initial step for mesh alignment was required. However, while (Marras et al., 2010) used parts of the same object from different meshes to fill in the holes, (Pauly et al., 2005) used parts of other objects. Because the objects are different, instead of aligning the meshes with an ICP type of algorithm, (Pauly et al., 2005) resorts to non rigid deformations. Both algorithms used the stitching algorithm proposed in (Turk and Levoy, 1994) to combine different meshes.

Other approaches to the stitching itself, but that assume that meshes were also previously registered are presented in (Borodin et al., 2002) and (Soucy and Laurendeau, 1995). The former algorithm stitches by introducing new edges and minimizes their length by creating and deleting vertices in the boundaries. In our algorithm, JASNOM, we also focus on minimizing the edge length, but we do it for the purpose of finding a rigid transformation that aligns the two meshes. The algorithm in (Soucy and Laurendeau, 1995) uses a Delaunay triangulation on a reprojection of non-overlapping meshes. However, the complete algorithm assumes that there is a very fine alignment between meshes.

JASNOM adds to the capabilities of the previous algorithms, the possibility of aligning meshes with no overlap and connecting meshes without resorting to existing triangles to ensure manifoldness.

3 Mesh Alignment

JASNOM addresses the problem of aligning and stitching two meshes M_1 and M_2 by focusing on the boundaries of each mesh, B_1 and B_2 as shown in Figure 3. In particular, JASNOM creates a complete mesh by assigning new edges from one boundary to the other and minimizing the total length of these edges by means of a rigid transformation. Furthermore, while minimizing edge length, it must prevent

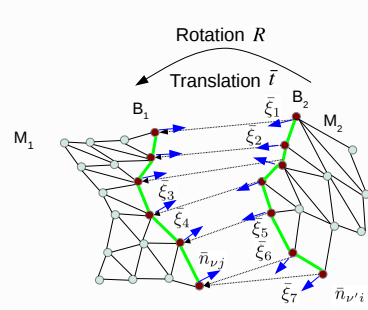


Figure 3: The two meshes are connected by assigning edges from one boundary to the other. We represent edges by error vectors ξ_i whose coordinates depend on R and \bar{t} . The smaller vectors, $\bar{n}_{v,v'}$, correspond to the boundary normals at the vertices.

the meshes from intersecting each other. Formally, JASNOM solves an optimization problem whose cost function, J , is composed of two independent terms J_1 and J_2 . The first term, J_1 , penalizes the total edge length, while J_2 penalizes the intersection. The result of the optimization is the mesh alignment and a initial set of assignments that will later be used for stitching.

3.1 Minimizing Edge Lengths

To ensure that edges are as small as possible, JASNOM addresses the aligning of two meshes as a registration problem, where edges represent assignments between vertices in the two meshes. These assignments are represented by a binary matrix A , whose element $A_{i,j}$ is equal to 1 if and only if vertex v_j in boundary B_1 is connected to vertex v'_j in boundary B_2 . Assuming there are K vertices in B_1 and N vertices in B_2 , $A \in \{0, 1\}^{K \times N}$ and if no additional constraints are added, there are $2^{K \times N}$ different assignment matrices.

Matrix A defines a set of error vectors, $\bar{\xi}_i$, each associated to a stitching edge. The error vector represents the displacement between assigned vertices in the two borders: $\bar{\xi}_i = \left(\sum_{j=1}^K A_{i,j} \bar{x}_j \right) - \bar{y}'_i$, where \bar{x}_j and \bar{y}'_i are the coordinates of the vectors in B_1 and B_2 in the same reference frame. However we only have access to the coordinates in the original reference frames, which differ by a rotation R and a translation \bar{t} . Therefore, the cost function J_1 , responsible for minimizing the length of the stitching edges, is given by Eq.1.

$$J_1(A, R, \bar{t}) = \sum_{i=1}^N \|\bar{\xi}_i\|^2 = \sum_{i=1}^N \left\| \left(\sum_{j=1}^K A_{i,j} \bar{x}_j \right) - R\bar{y}_i + \bar{t} \right\|^2 \quad (1)$$

3.2 Preventing Intersection

To globally ensure that no intersection occurred, JASNOM would have to check for local intersections between each and all the vertices in one mesh versus each and all the faces of the other mesh. JASNOM relaxes the problem by considering only intersections between a vertex $v'_i \in B_2$ and the neighborhood of $v_j \in B_1$ to which it was assigned.

Local intersections can be modeled by keeping track of the position of mesh $M_{1,2}$ with respect to each vertex of the boundary $B_{1,2}$. This relative position is represented for each boundary vertex v by the normal to the boundary \bar{n}_v , as shown in the Figure 3. Keeping in mind that error vectors $\bar{\xi}$ point from vertices $v' \in B_2$ to vertices $v \in B_1$, if $\bar{\xi}_i$ points in the opposite direction of $\bar{n}_{v'}$, the vertex $v'_i \in B_2$ is on top of mesh M_1 .

Ideally, preventing intersections would then result on a set of constraints in the optimization problem. However, since the estimation of the boundary normals is very sensible to noise and irregularities on the boundary, the constraints may yield the problem unsolvable. We thus relax these constraints by introducing them as a second term to the cost function, J_2 . The constraints are modeled as a sum of logistic functions that receive as argument the projection of $\bar{\xi}_k$ on $-\bar{n}_{v_k}$ and $\bar{n}_{v'_k}$ as in Eq.2. The logistic function penalizes edges that cross the opposite boundary by penalizing the negative projections on $\bar{n}_{v'_k}$ and the positive projections on \bar{n}_{v_k} .

$$J_2(A, R, \bar{r}, \lambda) = \sum_{k=1}^N \frac{1/N}{1 + \exp\{\lambda \bar{\xi}_k \cdot \bar{n}_{v'_k} / \|\bar{\xi}_k\|\}} + \sum_{k=1}^N \frac{1/N}{1 + \exp\{-\lambda \bar{\xi}_k \cdot \bar{n}_{v_k} / \|\bar{\xi}_k\|\}} \quad (2)$$

We introduce a slack variable λ to control the steepness of the logistic function. High values of λ correspond to steepest transitions on the logistic function and enforce the constraints more strictly. Lower values of lambda relax the constraints. The best value depends on the confidence on the normal estimation.

3.3 Minimizing the Cost Function

Formally, JASNOM aligns and stitches the two meshes by finding the matrices A and R and the vector \bar{r} that solve the optimization problem defined in Eq. 3

$$J(A, R, \bar{r}; \lambda, v) = J_1(A, R, \bar{r}) + \mu J_2(A, R, \bar{r}, \lambda) \quad (3)$$

$$s.t. \quad A \in \{0, 1\}^{N \times K}, \quad R \in O(3), \quad \bar{r} \in \mathbb{R}^3;$$

where $v \in \mathbb{R}^+$ weights the two cost functions and depends on the object or application. E.g., if the task is hole filling and the patch we use is smaller than the

hole there will be no intersection and thus μ can be set to zero.

Without further constraints, finding the matrix A is a combinatorial problem. However, we note that if the assignments between meshes correspond to edges in the mesh of an object, not all the assignments are valid. For example, no edge can cross the interior of the object. We explore the physical constraints in the problem to reduce the number of possible assignments between the two meshes. The constraints, which we address in Section 4, are independent of the rigid transformation that aligns the two meshes.

JASNOM is then able to tackle separately the discrete problem of finding the assignment matrix A from the problem of finding the rigid transformation, R and \bar{r} . The separation and reduced complexity allow the algorithm to address the discrete problem by enumeration, i.e., JASNOM minimizes $J(A, R, \bar{r})$ by finding the minimum over the set of all valid assignments, \mathcal{V}_A , using exhaustive search.

$$\arg \min_{A, R, \bar{r}} J(A, R, \bar{r}) = \arg \min_{A_\tau} \tilde{J}(R, \bar{r}; A_\tau) \quad (4)$$

$$s.t. \quad A_\tau \in \mathcal{V}_A$$

$$\tilde{J}(R, \bar{r}; A_\tau) = \min_{R, \bar{r}} J(A_\tau, R, \bar{r}) \quad (5)$$

The optimization problem expressed in Eq. 5 is non-convex. To find a local solution, we use a generic non-linear optimization algorithm, such as BFGS Quasi-Newton method (Broyden, 1970). To initialize the optimization, JASNOM first solves the relaxed problem obtained from Eq. 3 by setting $\mu = 0$, which has a closed form solution (Schönemann, 1966).

4 Valid Assignments

Stitching assignments in JASNOM correspond to edges in an object surface and, as shown in Figure 4(a.2), these edges have a specific geometric structure. In the following, we address the geometric properties that can be used to constrain possible assignments and then present how JASNOM uses the constraints to efficiently find the best stitching edges.

4.1 Assignment Constraints

The complete surface mesh of an object is an orientable 2-manifold mesh, while an isolated part of the surface is an orientable 2-manifold mesh with a boundary. In Figure 4(a.2) we exemplify the mesh structure corresponding to an object part. In particular, we note that there are only two types of edges:

those that belong to two triangles, and those that belong to only one, i.e., that are in the mesh boundary. Formal definitions of all these concepts can be found in computational geometry books, e.g., (Munkres, 1984). We briefly illustrate them here to allow a better comprehension of the constraints.

Object surfaces are orientable because they have an inside and an outside. Using one of these directions, it is possible to define consistently the normal directions for all points at the surface as shown in Figure 4(a). For 2-manifold meshes, the definition of a normal to a triangle is associated with a cyclic order of the triangle vertices. The normal to a triangle with vertices v_1, v_2 and v_3 with coordinates $\bar{x}_1, \bar{x}_2, \bar{x}_3 \in \mathbb{R}^3$ can be estimated by the outer product $\hat{n}_F = (\bar{x}_2 - \bar{x}_1) \times (\bar{x}_3 - \bar{x}_2)$. If the order of the vertices changes, the direction of the normal vector will be the exact opposite. To ensure consistency on the orientation of two adjacent faces, the two vertices of the common edge must be in opposite order, as shown in Figure 4(a.3). Boundary edges have only one possible orientation since they belong to a single triangle. This orientation defines the intrinsic direction of the boundary cycle, as shown in Figure 4(b).

The whole surface mesh is orientable if all adjacent faces are consistent. To guarantee that the union of two meshes is orientable, their boundaries cannot have a random orientation with respect to each other. JASNOM stitches two meshes by assigning an edge from one boundary to the other. This situation, illustrated in Figure 4(b.2), requires the orientation of the boundaries to oppose each other. This is in consistency with the Gluing theorem.

Since the union of the two meshes is introduced by the assignment matrix A , the matrix must reflect the ordering of the two boundaries. We thus introduce the constraint: $A_{i,j} = 1 \Rightarrow A_{i+1,j+k} = 0, \quad \forall k \geq 0$.

4.2 Order Preserving Assignments

The space of matrices that satisfy the previous constraint is still very large. To further constrain the valid assignments search space, \mathcal{V}_A , we introduce some geometric constraints. In particular, we note that if the two meshes were the exact complementary of each other over the object surface, the two boundaries would correspond to the same vertices and edges. In this case, given a mapping $\phi : B_2 \rightarrow B_1$ between the two boundaries that returns the point $v_j \in B_1$ equivalent to the point $v'_i \in B_2$, we can define the assignment between the two boundaries as $A_{i,j} = 1 \Leftrightarrow v_j = \phi(v'_i)$.

To construct this mapping, we define one origin in each boundary, and order the vertices according to

the boundary orientation. Assuming that the origins correspond to the same point, two points that are at the same distance, i , from the origin, should be equivalent to each other. To account for the opposite boundary orientations, the mapping needs to invert the vertex ordering, e.g., as in $\phi(v'_i) = v_{N-i}$. This is illustrated in Figure 5 where N refers to the total number of vertices in the boundary and i to the order of the vertex v'_i with respect to the boundary of B_2 .

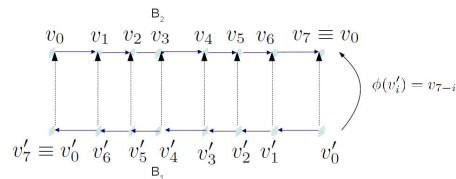


Figure 5: Example for the construction of an assignment between boundaries in the limit case where the vertices in both boundaries coincide exactly.

For the vertices of the two boundaries to map to each other, the sampling in both surfaces has to be exactly the same. Thus, in most cases, mapping the vertices order across boundaries does not preserve the object geometry. It is then more reasonable to map distances over the boundaries. In this work, we use the normalized curve length $l \in [0, 1]$ to account for those cases when the boundaries do not have the exact same length. In this case, the previous map can be rewritten as $\phi(l') = 1 - l$.

After mapping a point between boundaries based on the normalized length, JASNOM still needs to find the closest vertex to that point. This search can be efficiently implemented by introducing an ordering function $f(l) : [0, 1] \rightarrow [0, N]$, which maps lengths over a specific boundary to a vertex order. For example, if vertex v_k is at a length l_k , $f(l_k) = k$. For values of l that do not correspond to exact vertices length but to points on the boundary edges, $f(l)$ returns the order of the closest vertex.

Using the map $\phi(l')$ and knowing the ordering function $f(l)$ for B_1 , we can find the order j of the vertex $v_j \in B_1$ to which assign $v'_i \in B_2$ by performing three steps. Namely:

- i) computing the length $l'_i = l'(v'_i) = l'_{i-1} + \|\bar{y}_i - \bar{y}_{i-1}\|$;
- ii) mapping the length l'_i to the length l of the equivalent point in B_1 : $l = \phi(l'(v'_i))$;
- iii) finding the vertices in B_1 that have a distance to the boundary closest to l using the ordering function over B_1 : $j = f(\phi(l'(v'_i)))$.

The three steps are illustrated in Figure 6.

By repeating for all $v_i \in B_2$, JASNOM defines the

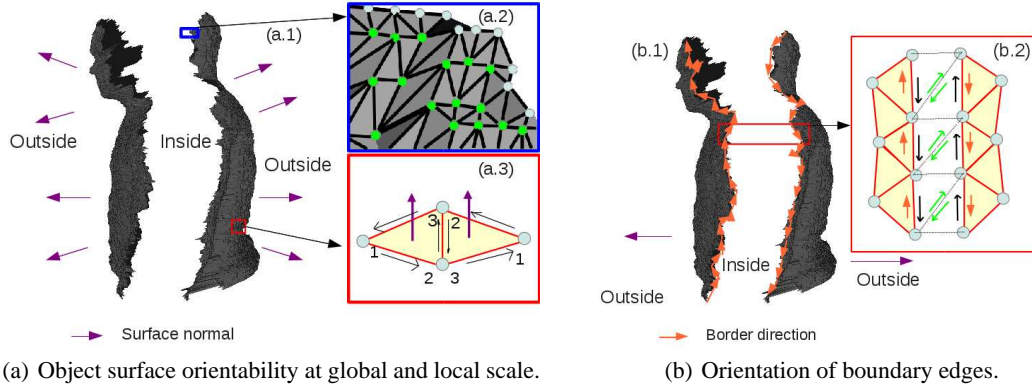


Figure 4: Order constraints on the boundary vertices. The image on the left shows how the orientability of an object induces an ordering at the edge level. The image on the right shows how the ordering is reflected in the boundary and how it constrains the assignments between boundaries. See the text for further explanations.

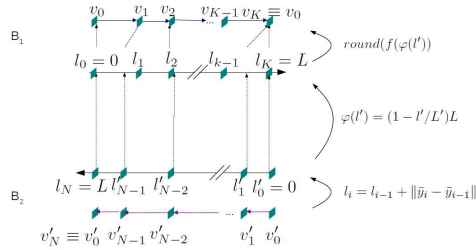


Figure 6: Three steps approach to define order preserving assignments between the boundaries. See text for details.

assignment matrix A as

$$A_{i,j} = 1 \Leftrightarrow j = \text{round}(f(\varphi(l'(v_i)))) \quad (6)$$

The previous definition for A depends only on the map φ and the ordering function $f(l)$. However, both functions depend on the vertex defined as an origin on either boundary. If any other vertex $v'_\tau \in B_2$ was assumed to be equivalent to the origin, $v_0 \in B_1$, the mapping could be recovered by shifting l' by l_τ . This origin ambiguity is translated into N different valid maps between boundaries.

JASNOM addresses the ambiguity problem by considering all possible N different shifts τ of the boundary B_2 with respect to the boundary B_1 . Each shift gives rise to a new mapping φ_τ and each mapping gives rise to a new assignment matrix A_τ . Thus, the combinatorial problem can be reduced into N independent problems. We note that by changing the shift in B_2 and not in B_1 , the ordering function defined in B_1 will be the same in all the shifts in A_τ .

5 Final Stitching

After aligning both meshes, JASNOM uses the best assignment to reconstruct the manifold M_c . In partic-

ular, the assignment as defined in 6 ensures that each vertex in B_2 already has an edge connecting it to a vertex in B_1 . However, not all the vertices in B_1 have an edge connecting to B_2 and some vertices in B_1 have more than one edge. Furthermore, just ensuring that there is an edge for all the vertices, does not ensure that the end result is a triangular mesh.

To stitch the meshes together, we use two simple strategies. First, we create a triangular mesh from the assignments already present. Then we assign the missing edges on B_1 so that they do not cross the edges already present.

For the first step, JASNOM adds a second edge to all the vertices $v'_i \in B_2$. As shown in Figure 7(b), the target vertex, $v_i \in B_1$ of the second edge of v'_i is the first target of the next vertex, $v'_{i+1} \in B_2$.

In the second step JASNOM assigns the missing edges in B_1 by running through all the vertices $v_i \in B_1$ by their reverse order. As shown in Figure 7(c) each vertex with no edge is assigned the same target vertex $v'_i \in B_2$ as the target of the previous vertex $v_{j-1} \in B_1$.

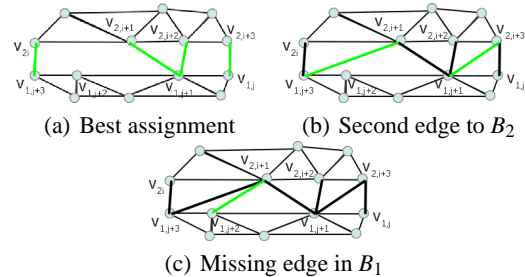


Figure 7: Schematic for the stitching between the two meshes given the set of one to one correspondences that result from the alignment stage. See the text for comments.

This strategy locally ensures manifoldness since there are no crossings between neighboring edges. The constraints in the assignments ensure that the ini-

tial set of edges do not cross and the new edges always preserve the ordering between boundaries.

In summary, JASNOM creates a complete 3D object surface model from non overlapping meshes by enumerating all valid assignment matrices, $A_\tau \in \mathcal{V}_A$ and, for each matrix, finding the rigid transformation that minimizes the cost function $J(A_\tau, R, \vec{t})$. JASNOM chooses the best assignment as the one that minimizes the cost function over all the minima, and aligns the meshes accordingly. This assignment serves also as initialization to the stitching algorithm, where the missing triangles are added.

6 Proof of Concept

We test our stitching algorithm with three experiments. In the first we illustrate its potential for fast 3D object scanning by modeling two smooth objects. In the second, we illustrate its potential for reconstructing 3D models from articulated objects such as humans. Finally, in the third experiment, we illustrate its potential for hole filling.

For the first experiment, we model two objects. The first is the electric pitcher, Figure 1, and the second a book, Figure 8. To collect both meshes for the example, we retrieve an image with the object in its regular position and then flip it upside down to collect the second image. The complete process is extremely fast from a user perspective and does not require previous registration of multiple cameras. The resulting complete meshes are presented in Figure 9.

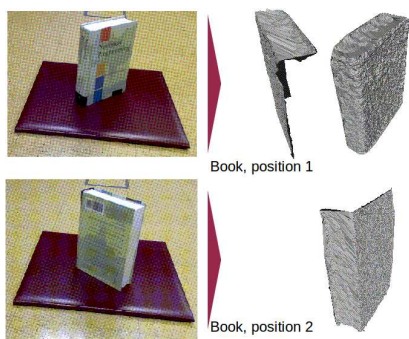


Figure 8: Acquisition of two meshes from a book.

For the purpose of accuracy while estimating centroids and other intermediate steps, JASNOM interpolates boundaries to ensure an uniform and dense distribution of points. To deal with the non-compactness of the object, JASNOM selected just the longest boundary. We note that the reconstructed objects show a good match at the boundaries.

For the second experiment, two range images of

the upper body of a human were retrieved simultaneously by two unregistered Kinect cameras. The complete mesh obtained with JASNOM algorithm is shown in Figure 10. We note that the two meshes do not cover the complete object and there are several large missing parts across the boundary. However, by preventing intersection, JASNOM was able to keep the overall human structure. In particular, the hole created by the cut at the waist is large enough that by simply attempting to minimize the distance between points, would lead to mesh intersections. Again we note that, with no previous camera registration, JASNOM created a rough shape of a non-rigid object using two Kinect cameras.

For the last experiment, we use a simple range image of an object with a hole and a small patch retrieved from another mesh, Figure 11(a). JASNOM covered and stitched the hole, Figure 11(b). Since the objective is to insert the patch on the hole in the other mesh, we did not penalize intersections between meshes, i.e., $\mu = 0$. We note that in this case the re-triangulation method left a smooth surface after patching the hole.

7 Conclusions

In this work we have contributed an algorithm, JASNOM, that allows the joint alignment and stitching of non-overlapping meshes, and provided evidence of its potential for fast 3D object scanning through simple experiments with data obtained with a Kinect camera.

From the experiments we conclude that JASNOM successfully constructs 3D models of different object types, including rigid and non rigid. The success of JASNOM is due mostly to the cost function definition. By preventing the intersection between boundaries, JASNOM preserves the object structure even with noisy boundaries. JASNOM is thus able to reconstruct complex shapes with missing parts such as the human we presented.

When compared with existing stitching algorithms, JASNOM adds the capability to create complete models without previous registration of individ-

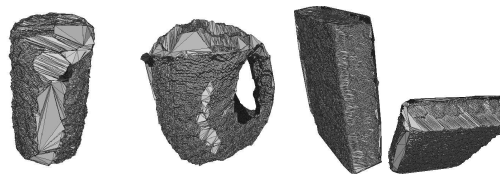


Figure 9: Reconstruction of man made objects using JASNOM. The first row presents two different views from the electric pitcher and the second from the book.

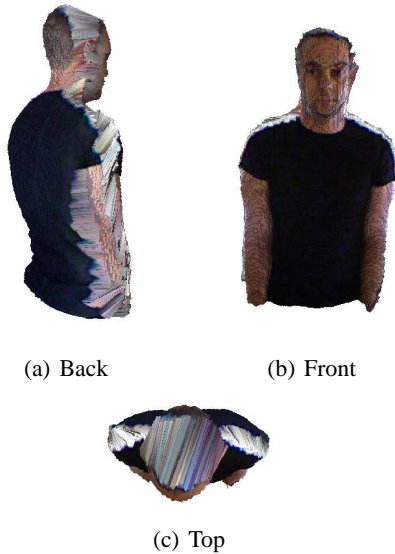


Figure 10: Human model completed using JASNOM.

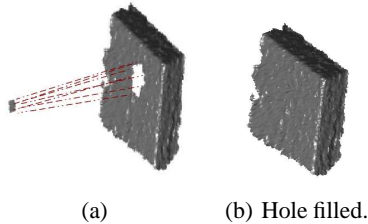


Figure 11: Results for the hole patching experiment using JASNOM. Figure 11(a) presents the original mesh with a hole and the patch. Figure 11(b) presents the glued mesh.

ual meshes. The registration typically requires overlap between the two meshes, which is not always available or convenient. JASNOM also does not require the calibration of one camera position with respect to the other. The registration and construction of models can be easily achieved with little effort and setup preparation. This allows for the fast creation of extensive 3D (possibly 3D+RGB) models data sets.

JASNOM assumes that the two meshes are complementary over the object surface and, while we showed it could reconstruct objects in more general cases, e.g., the human shape, other objects might not be reconstructed so easily. In particular, we note that the boundaries of the human shape meshes, had a preferential direction, i.e., the elongated shape means that small deviations from the best assignment between boundaries lead to steep increases in the cost functions. More symmetric objects do not benefit from the steepness in the cost function and the alignment is more sensitive to gaps between boundaries. A possible approach, which we will explore in future

work, is to reintroduce the asymmetries by penalizing color discontinuities at the boundaries.

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