

**Supplemental Information**  
**for**  
***Spin Transformations of Discrete Surfaces***

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## 1 Performance

The main computational cost when computing conformal deformations is solving the eigenvalue problem

$$(D - \rho)\lambda = \gamma\lambda$$

for the smallest eigenvalue  $\gamma$  and corresponding eigenvector  $\lambda$ . In practice, however, **a good solution can often be found by solving a single linear system.**

More specifically, let  $X = A^*A$  be the matrix used in the standard eigenvalue problem. This matrix is symmetric and positive-semidefinite with only about 7 nonzeros in each row/column, which means it can be efficiently inverted using standard methods like Cholesky factorization or the conjugate gradient method. A simple algorithm for computing the smallest eigenvector is then:

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**Algorithm 1** The Inverse Power Method

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**Require:** Initial guess  $\lambda_0$ .

- 1: **for**  $i = 1, \dots, k$  **do**
  - 2:   Solve  $X\lambda_i = \lambda_{i-1}$
  - 3:    $\lambda_i \leftarrow \lambda_i / \|\lambda_i\|$
  - 4: **end for**
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Starting from a random initial guess this procedure tends to produce good results after only a few iterations (see Figure 1). A more intelligent initial guess sets  $\lambda = 1$  at each vertex, which corresponds to the identity transformation. In this case we often get *very* close to the solution after only a single iteration, requiring only a single linear solve (see Figure 2).

The figures below depict two tests: *bumpy*, where we add random bumps to a sphere, and *moon*, where we paint a face on a disk. On a 2.4 GHz Core 2 Duo laptop, a single iteration (using `mldivide` in MATLAB) takes 0.3 seconds on a mesh with 8k faces (*bumpy*) and 1.26 seconds on a mesh with 33k faces (*moon*).

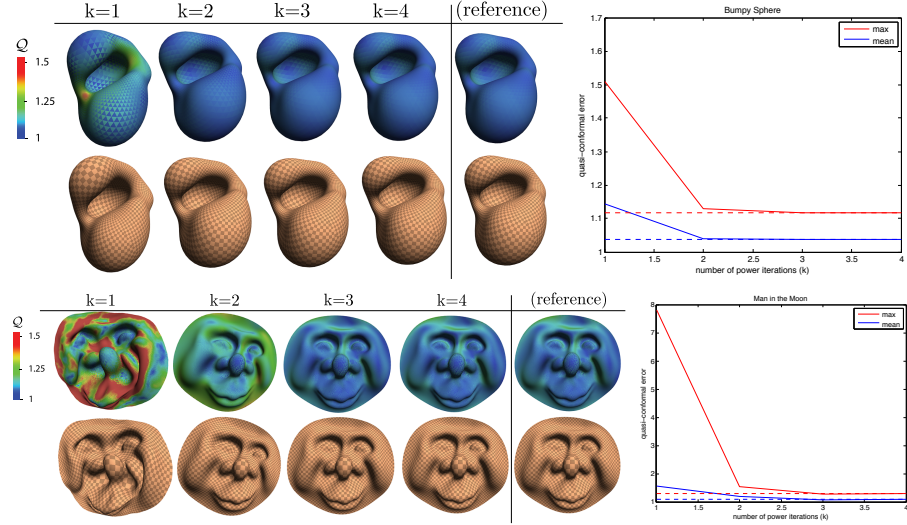


Figure 1: *Left:* solutions for bumpy and moon found after  $k$  iterations of the inverse power method. *Right:* quasi-conformal distortion as a function of iteration count. A random initial guess was used in both tests.

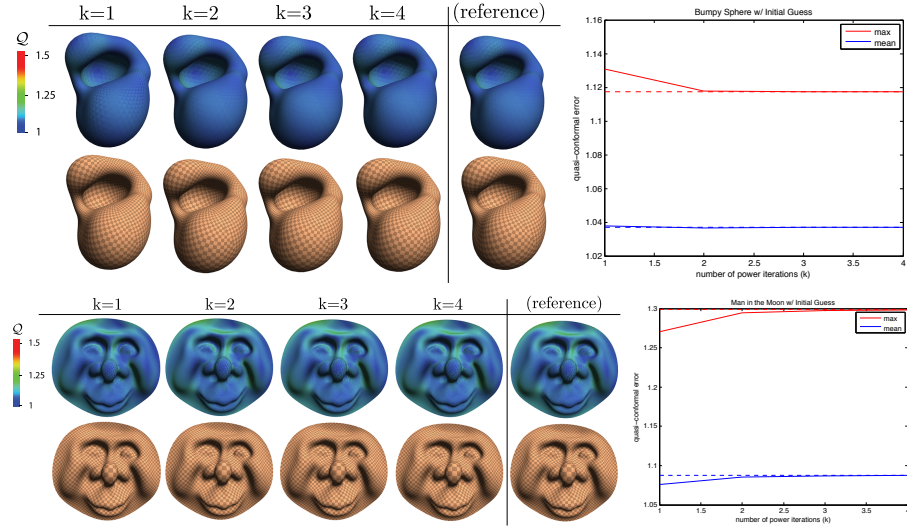


Figure 2: The same test depicted in Figure 1 but using a constant initial guess. Note that after the first iteration we obtain a solution that is virtually indistinguishable from the reference solution.

## 2 Further Performance Enhancements

Suppose a good initial guess  $\lambda_0$  is known (e.g.,  $\lambda_0 = 1$ ). Then the Rayleigh quotient

$$\frac{\lambda_0^H \mathbf{X} \lambda_0}{\lambda_0^H \lambda_0}$$

should provide a good estimate  $\gamma_0$  of the smallest eigenvalue. One can therefore improve the rate of convergence of the inverse power method by using the shifted matrix  $\mathbf{X}' = \mathbf{X} - \gamma_0 \mathbf{I}$  in place of the usual matrix  $\mathbf{X}$  (where  $\mathbf{I}$  is the identity).

## 3 Building the Eigenvalue System

In practice it may be convenient to build the matrix  $\mathbf{X}$  directly (rather than building it up from constituent matrices). The following algorithm provides a simple *facewise* construction of  $\mathbf{X}$ :

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### Algorithm 2 Facewise Construction of Eigenvalue System

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```

1: for  $k = 1, \dots, |F|$  do
2:    $a \leftarrow -\frac{1}{4\mathcal{A}}$ 
3:    $b \leftarrow \rho/6$ 
4:    $c \leftarrow \rho^2 \mathcal{A}/9$ 
5:   for all  $(i, j) \in \{1, 2, 3\} \times \{1, 2, 3\}$  do
6:      $\mathbf{X}_{ij} \mathrel{+}= a e_i e_j + b(e_j - e_i) + c$ 
7:   end for
8: end for
```

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Here  $k$  is the index of the current face,  $\mathcal{A}$  is its area, and  $\rho$  is the desired change in mean curvature half-density. The inner loop visits *all* ordered pairs of edges  $e_1, e_2, e_3$  of the current face. In C++, the algorithm might look something like Listing 1.

**Listing 1: Facewise Construction of Eigenvalue System in C++**

```
void buildEigenvalueProblem( const vector<Face>& faces,
                             const vector<Vertex>& vertices,
                             QuaternionSparseMatrix& E )
{
    // allocate a sparse |V| $\times$ |V| matrix
    int nV = vertices.size();
    E.resize( nV, nV );

    // visit each face
    for( size_t k = 0; k < faces.size(); k++ )
    {
        double A = face[k].area();
        double rho = face[k].rho;

        // compute coefficients
        double a = -1. / (4.*A);
        double b = rho / 6.;
        double c = A*rho*rho / 9.;

        // get vertex indices
        int I[3] =
        {
            faces[k].vertex[0],
            faces[k].vertex[1],
            faces[k].vertex[2]
        };

        // compute edges across from each vertex
        Quaternion e[3];
        for( int i = 0; i < 3; i++ )
        {
            e[i] = vertices[ I[ (i+2) % 3 ] ] -
                    vertices[ I[ (i+1) % 3 ] ] ;
        }

        // increment matrix entry for each ordered pair of vertices
        for( int i = 0; i < 3; i++ )
        for( int j = 0; j < 3; j++ )
        {
            E(I[i],I[j]) += a*e[i]*e[j] + b*(e[j]-e[i]) + c;
        }
    }
}
```

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