HW5 Question 4

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1 HKS calculation (see Fig. 1)

• Using the HKS as the mesh descriptor function, e.g F and G in the lecture 22.

```
// First get the function representation of the source (A) and target (B)

MatrixXd F = load_csv(root_path + "Results/" + source + "_hks.csv", 6890, 100);//dim of the des
MatrixXd G = load_csv(root_path + "Results/" + target + "_hks.csv", 6890, 100);
```

Figure 1: Compute the priority of a halfedge.

• Get the Pseudo inverse of bases (see Fig.2)

```
MatrixXd S_eigval = load_csv(root_path + "Results/" + source + "_eigval.csv", 300, 1);
MatrixXd T_eigval = load_csv(root_path + "Results/" + target + "_eigval.csv", 300, 1);

Map<MatrixXd, 0, OuterStride<>> s_base(S. data(), S.rows(), K, OuterStride<>(S. outerStride()));//6890*40

Map<MatrixXd, 0, OuterStride<>> t_base(T. data(), T.rows(), K, OuterStride<>(T. outerStride()));//6890*40

// Get the pseudo inverse of the bases

MatrixXd trans_m_bases = s_base.completeOrthogonalDecomposition().pseudoInverse();//40*6890

MatrixXd trans_m_bases = t_base.completeOrthogonalDecomposition().pseudoInverse();//40*6890
```

Figure 2: Pseudo inverse of the bases.

• Get the A&B representation (see Fig. 3

2 Least Square Fitting.

Implement the Least Square Regression of two functions (See Fig. 4).

```
Map<VectorXd, 0, OuterStride<> > s_diag(S_eigval.data(), K, S_eigval.cols(), OuterStride<>(S.outerStride()));
Map<VectorXd, 0, OuterStride<> > t_diag(T_eigval.data(), K, T_eigval.rows(), OuterStride<>(T.outerStride()));
MatrixXd A = trans_m_bases * F;//40*100
MatrixXd B = trans_n_bases * G;//40*100
```

Figure 3: Get the representation of A and B.

```
tuple<MatrixXd, float> solver(MatrixXd p1, MatrixXd p2, VectorXd L1, VectorXd L2, float lambda)
    //implement the fitting of ||C*p1-p2||+1ambda*||C*L1-L2*C||
    int n1 = p1. rows();
   int n2 = p2. rows();
   MatrixXd A_fixed = p1 * p1. transpose();
    MatrixXd B = p1 * p2. transpose();
    MatrixXd X = MatrixXd::Zero(n2, n1);
    for (int i = 0; i < n2; i++)
        VectorXd test = lambda * (L1.array() - L2(i)).square();
       Eigen::MatrixXd diag = test.array().matrix().asDiagonal();
       diag += A_fixed;
        VectorXd sol(Map<VectorXd>(B. col(i).data(), B. rows()));
       X. row(i) = diag.bdcSvd(ComputeThinU | ComputeThinV).solve(sol);
   float residual = ((X*p1) - p2).norm();
   residual *= residual;
   return make_tuple(X, residual);
```

Figure 4: Least Square Fitting of two terms.

3 Point to Point match (see Fig. 5)

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4 Implementation and Results.

- You can implement all meshes' FM calculation and matching by running Report_total_error() function (in func_map.cpp).
- Evaluate the FM matching (Achieve mean error:0.6950, see Fig. 6).

Figure 5: FM matching with the projector.

```
Mean error of source mesh 1: 0.7395207031202279
Mean error of source mesh 2: 0.5369801875309134
Mean error of source mesh 3: 0.6241344359691187
Mean error of source mesh 4: 0.749525471690207
Mean error of source mesh 5: 0.7013338146651498
Mean error of source mesh 6: 0.6630037559967478
Mean error of source mesh 7: 0.6925111184599908
Mean error of source mesh 8: 0.761538572759552
Mean error of source mesh 9: 0.7869854944931645
Mean error of all source meshes: 0.6950592838538968

Process finished with exit code 0
```

Figure 6: Result of FM matching (when setting $\lambda = 5$).